

**DEVELOPMENT, CHARACTERIZATION AND OPTIMIZATION OF
ELECTROLESS Ni-W-P-Ce COATINGS FOR MULTIFUNCTIONAL
APPLICATIONS**

Thesis submitted by

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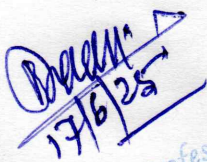
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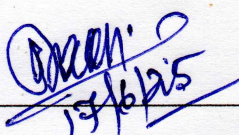
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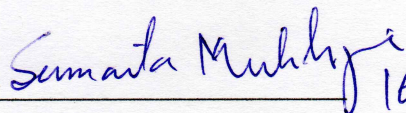
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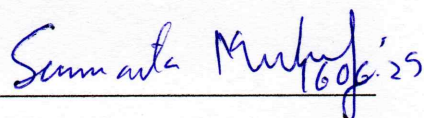
This is to certify that the thesis entitled “**Development, characterization and optimization of electroless Ni-W-P-Ce coatings for multifunctional applications**” submitted by Shri VIKASH KUMAR, who got his name registered on 21.01.2021 for the award of Ph.D. (Engineering) degree of Jadavpur University is absolutely based upon his own work under our supervision and that neither his thesis nor any part of the thesis has been submitted for any degree/diploma or any other academic award anywhere before.

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ABSTRACT

Electroless coating is a chemical deposition technique that allows for uniform application of metal or alloy coatings onto a wide range of substrates, both metallic and non-metallic, regardless of their shape or complexity. Unlike electroplating, it operates without the need for an external electric current and relies on controlled redox reactions to reduce metal ions from the solution onto the substrate. This thesis focuses on the development, characterization, and optimization of electroless Ni-W-P-Ce coatings on low-carbon steel (AISI 1010) for multifunctional applications.

For development of the coating, different coating conditions were explored, and a neutral bath (pH 7.5) stabilized with thiourea was identified to exhibit superior performance in terms of bath stability, deposition rate, coating microstructure, and microhardness. Next, the bath composition was optimized to maximize the overall coating performance, and a novel hybrid multi-objective optimization approach was adapted by combining an artificial neural network (ANN) with a metaheuristic algorithm. Among the evaluated algorithms, red fox optimization (RFO) proved most effective, yielding an optimized coating with a surface roughness of 0.12 μm , microhardness of 726 HV, 64% reduced fouling weight, surface free energy of 15.97 J/m², and 98.73% corrosion protection efficiency.

Next, a comprehensive suite of characterization techniques, including microhardness, surface roughness, deposition rate, scratch and wear resistance, corrosion resistance, biofilm formation, surface morphology, structural analysis, and compositional analysis, were employed to thoroughly investigate the influence of cerium incorporation in Ni-W-P-Ce coatings. An optimal Ce concentration of 8 mg/L was identified, which significantly enhanced the functional performance of the coatings. Subsequent annealing at various temperatures revealed a trade-off among the properties, indicating that careful thermal treatment can balance the mechanical, chemical, and structural attributes. Further studies explored the effects of substrate surface preparation methods, including activation with different agents (HCl and PdCl₂), polishing to various surface roughness levels, and modifying bath loading conditions. Among these, HCl activation was more effective than PdCl₂ activation, enhancing adhesion through micro-roughening. Polishing with finer grit (1500 grit) promoted controlled nucleation and uniform coating growth, resulting in superior adhesion and enhanced mechanical properties. Additionally, an optimal bath loading of 0.88 dm²·L⁻¹ was identified, effectively balancing the deposition rate and bath stability.

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ABBREVIATIONS

ENC	Electroless nickel coating
CVD	Chemical vapor deposition
PVD	Physical vapor deposition
HER	Hydrogen evolution reaction
PCB	Printed circuit board
DoE	Design of experiments
RSM	Response surface methodology
ANN	Artificial neural network
GA	Genetic Algorithm
ABC	Artificial bee colony
FA	Firefly algorithm
TLBO	Teaching learning-based optimization
WOA	Whale optimization algorithm
RFO	Red fox optimization
UBM	Under bump metallization
REEs	Rare earth elements
CCD	Central composite design
BBD	Box-Behnken design
ANOVA	Analysis of variance
BPNN	Back-propagation neural network
AHA	Artificial hummingbird algorithm
SA	Simulated annealing
NSGA	Non-dominated sorting genetic algorithm
MWCNT	Multi-walled carbon nanotube

AI	Artificial intelligence
ML	Machine learning
UTS	Ultimate tensile strength
FESEM	Field-emission scanning electron microscopy
EDS	Energy-dispersive X-ray spectroscopy
XRD	X-ray diffraction
TDS	Total dissolved solids
OCP	Open circuit potential
LP	Linear potentiodynamic polarization
HV	Microhardness
FW	Fouling weight
SFE	Surface free energy
PE	Protection efficiency
MSE	Mean squared error
FWHM	Full width at half maximum
EIS	Electrochemical impedance spectroscopy
OHL	Outer Helmholtz layer
IHL	Inner Helmholtz layer
BE	Binding energy
EG	Ethylene glycol
SFC	Scratch friction coefficient
SV	Scratch volume
CoF	Coefficient of friction
SWR	Specific wear rate
CPE	Constant phase element
FTIR	Fourier transform infrared spectroscopy

Chapter 1 Introduction

This chapter provides an in-depth overview of electroless coatings and covers their principles, types, industrial applications, and emerging research trends. This highlights the increasing demand for multifunctional coatings and the development of electroless polyalloy coatings on low-carbon steels and concludes with a summary of the thesis structure and research objectives.

1.1 Background

Surface modification techniques play a crucial role in enhancing the properties of materials, with applications spanning multiple industries. These techniques can be broadly classified into two categories: coating and surface treatment. Coatings involve the application of thin layers to improve corrosion resistance, wear resistance, and thermal stability. In contrast, surface treatments, such as diffusion and plasma treatments, modify surfaces without adding layers, thereby enhancing the mechanical and chemical properties of the material at the surface level [1].

Coating techniques can be classified based on the state of the coating material during the deposition (Figure 1.1). Gaseous functional coating methods such as chemical vapor deposition (CVD) and physical vapor deposition (PVD) are vacuum-based processes that form thin films with enhanced mechanical and optical properties. These methods modify the near-surface structure and improve the performance in various applications.

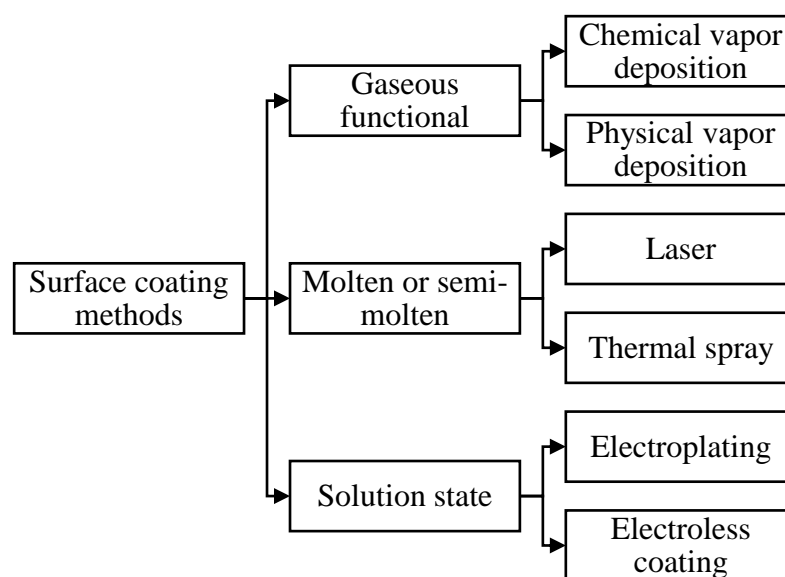


Figure 1.1. Classification of coating techniques based on deposition state.

Molten or semi-molten techniques, such as laser cladding and thermal spraying, involve heating materials until they reach a molten state before deposition onto a substrate. Although these processes offer excellent thermal and wear resistance, their effectiveness is sometimes limited because of their poor adhesion properties. Solution-based coating techniques, including electroplating and electroless coating, are cost-effective and versatile [2]. In this study, electroless coatings are investigated.

1.2 Electroless coatings

Among the various coating methods, electroless coatings have garnered significant attention because of their ability to uniformly deposit metal layers via autocatalytic chemical reactions. Electroless coatings are chemical deposition processes that allow for uniform application of metal or compound coatings onto various substrates without the use of an external electric current. This method is valued for its versatility, as it can effectively coat both conductive and non-conductive materials, making it ideal for diverse applications including electronics, energy storage devices, and anti-corrosion solutions. The advantages of electroless deposition include the ability to produce high-quality, consistent coatings with exceptional anti-corrosion properties and excellent wear resistance, while avoiding issues related to current density uniformity typical of electroplating. However, this method also has some disadvantages, such as a limited bath life that requires regular monitoring and replenishment, potential environmental concerns associated with the disposal of alkaline and acidic baths, and the necessity for surface pre-treatments to ensure adhesion on certain substrates [3]. The best part of electroless coating is that it can be performed professionally in large batches, making the entire process more cost-effective and generating significant long-term savings.

Electroless coating is an autocatalytic process, in which the deposited metal acts as a catalyst for continuous deposition, creating a self-sustaining reaction. The process begins with the reduction of metal ions in a solution containing a suitable reducing agent that donates electrons to form a solid metallic layer on the substrate surface. In an electrochemical process, an active catalytic surface is required to facilitate the oxidation and reduction reactions that are essential for metal deposition. The effectiveness of deposition depends on the compatibility between the metal and the reducing agent, as some metals may not react efficiently, leading to incomplete deposition or side reactions. For instance, Pd exhibits a high catalytic activity across various reducing agents, making it

versatile under different deposition conditions. Similarly, copper effectively catalyzes formaldehyde oxidation, whereas nickel and cobalt can oxidize a broader range of reducing agents but are ineffective for formaldehyde oxidation [3–5].

Electroless deposition offers several advantages over traditional coating methods such as electroplating, PVD, and CVD. Its cost-effectiveness stems from the use of simpler and less expensive equipment and materials compared to electroplating, which requires precise control of the electrical parameters [6,7]. This makes electroless deposition ideal for large-scale applications, where the cost is a concern.

A key advantage is its ability to produce uniform coatings, even in complex geometries, owing to its autocatalytic nature [8]. Unlike electroplating, which struggles with uniform thickness on intricate parts, electroless deposition ensures consistent coverage without requiring a conductive substrate [9]. In addition, it is highly versatile, allowing coatings to be deposited on metals, polymers, and ceramics, unlike PVD and CVD, which have substrate limitations [10,11]. This broad compatibility expands its industrial application and requires less specialized equipment and expertise [12,13].

Electroless deposition also offers environmental benefits because certain formulations use less hazardous reducing agents and produce less toxic waste than electroplating, which often involves harmful chemicals, such as cyanides [14]. The development of sugar-based coating solutions further highlights their potential for sustainable manufacturing [6]. As industries prioritize cost, uniformity, and eco-friendliness, electroless deposition continues to be a valuable surface engineering technology.

Electroless coatings primarily consist of Ni-based alloys, particularly Ni-P and Ni-B coatings. Ni-P coatings are widely recognized for their exceptional corrosion resistance, which varies depending on the phosphorus content. A higher phosphorus content enhanced the corrosion resistance, whereas a lower phosphorus content increased the hardness. Conversely, Ni-B coatings exhibit superior hardness and wear resistance, but lower corrosion protection than Ni-P coatings. Furthermore, electroless composite coatings incorporating micro- and nanoscale reinforcements, such as SiC and TiO₂, significantly enhance the surface properties, including hardness, wear resistance, and lubricity [2].

1.2.1 Application of electroless coatings

Electroless coatings are extensively used industrially because of their durability and superior performance in harsh environments. As depicted in Figure 1.2, these coatings have applications in the automotive and electronics sectors, accounting for approximately 20% of the total usage. The oil and chemical (15%), machinery (11%), and aerospace (10%) industries rely heavily on electroless coatings for their corrosion and wear resistance. Other specialized industries, including food processing and printing, also benefit from electroless coatings, owing to their hygienic and protective properties [2]. Recent advancements in electroless coating technology include the development of a 24/7 helpline for troubleshooting and technical support, thereby highlighting the growing significance of this coating technique in industrial applications [15].

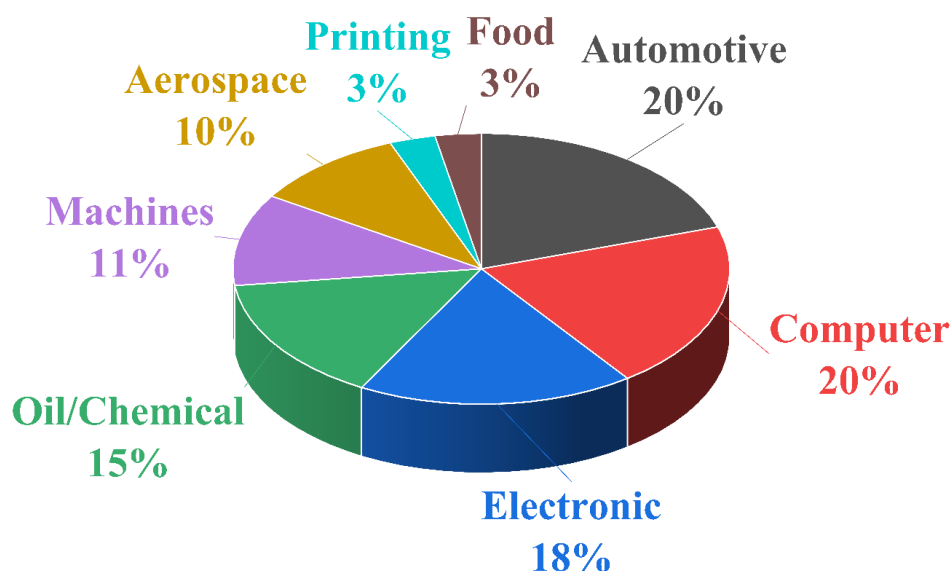


Figure 1.2. Industrial applications of electroless coatings across various sectors.

Electroless nickel coatings (ENC) have been proven to significantly extend the service life of components for up to 30 years in some cases and are extensively used across multiple industries. In power electronics, ENC is applied to protect cold copper plates and heat sinks, whereas in aerospace, oil and gas processing, and hydraulic systems, it safeguards actuator bodies and critical equipment subjected to extreme conditions. The automotive sector benefits from ENC by improving the durability and corrosion resistance of brake pistons, as well as preventing galvanic corrosion in electric vehicle busbars, thereby ensuring stable performance in hybrid and electric vehicles. The most commonly coated industrial components are illustrated in Figure 1.3 [16–22]. Notably, electroless coatings, such as Ni-

B and Ni-P/B, play a vital role in enhancing the durability and performance of components such as gears, bearings, valves, carburetors, and fuel injection systems, especially in environments involving alternative fuels [23,24]. Their excellent friction-reducing and wear-resistant properties extend the lifespan of various engine components [25,26].



Figure 1.3. Common industrial components benefiting from ENC.

Aerospace applications also benefit significantly because electroless coatings provide protective finishes for landing gear components, engine mounts, and hydraulic fittings, ensuring reliability under harsh environmental conditions [24,27]. Additionally, these coatings help to achieve lightweight and corrosion-resistant surfaces, which are essential for improving aircraft performance and safety [25].

In the electronics industry, electroless nickel and palladium coatings are widely applied to printed circuit boards (PCBs) to enhance their solderability and corrosion resistance [25,28]. Electroless silver coatings are used to improve the conductivity of circuit interconnects, whereas nickel coatings contribute to semiconductor manufacturing and metallization of plastics, enabling miniaturization and intricate designs [27,29].

Biomedical applications of electroless coatings include their use in medical implants and devices, where biocompatible surfaces facilitate better adhesion and functionality

[27,30]. Furthermore, these coatings play a crucial role in manufacturing composites and magnetic materials, offering superior mechanical strength and thermal conductivity [31].

Another significant advantage of electroless coatings is their ability to provide uniform metal deposition on complex geometries and non-conductive substrates. This makes them indispensable for a wide range of industrial applications, including solid free-form fabricated components, such as prosthetic limbs, which require both performance and aesthetic qualities [3,32].

One of the most significant recent applications of electroless coatings is their role as electrocatalysts in hydrogen evolution reactions (HER). The unique microstructures of these coatings, which often feature amorphous phases and high surface roughness, enhance their catalytic activities. Their superior corrosion resistance, favorable hydrogen adsorption energy, and optimized surface morphology contributed to their efficient and stable HER performance. The introduction of dopants, multilayer coatings, and process optimization further improve catalytic efficiency, making electroless nickel coatings a promising alternative to platinum-based catalysts for hydrogen production [33].

1.2.2 Types of electroless nickel coatings

As shown in Figure 1.4, there are several distinct types of electroless coatings, each designed to deliver specific performance benefits and characteristics that are suitable for a wide range of applications [34,35].

Pure nickel coatings

Pure electroless nickel coatings consist primarily of nickel, without significant additions of other elements such as phosphorus or boron. Hydrazine is commonly used as the reducing agent during deposition. Although pure nickel coatings tend to be softer and less corrosion-resistant than nickel-phosphorus alloys, they still provide excellent electrical conductivity and ductility. These properties make them suitable for specific applications in which conductivity and flexibility are essential [36].

Nickel alloy coatings

Nickel alloy coatings involve the addition of other metals (such as P, B, Zn, Ce, Cu, and W) to nickel, thereby improving the mechanical properties of the coating. These coatings typically exhibit enhanced hardness and wear resistance compared with those of pure

nickel. For instance, nickel-phosphorus and nickel-boron alloys are common variations with distinct attributes. The specific alloy composition can be tailored according to the requirements of the application, providing flexibility in properties, such as thermal stability and electrical conductivity. Variations in the levels of phosphorous and boron can lead to different categorizations.

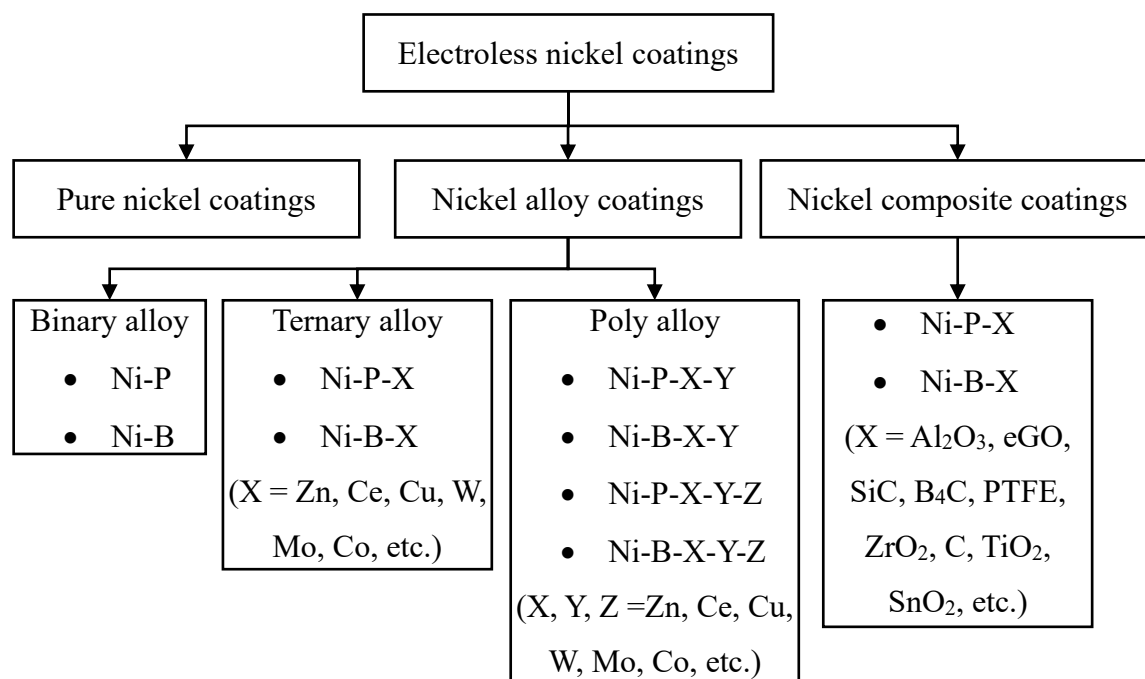


Figure 1.4. Classification of electroless nickel coatings.

High-P coatings (10-13% P) offer superior corrosion resistance, making them ideal for highly corrosive environments. Mid-phosphorus coatings (5-9% P) provide a balance between corrosion and wear resistance while delivering a bright finish, which is commonly used for moving parts, although they may induce tensile stress in certain metals, such as steel. Low-phosphorus coatings (up to 5% P) excel in high-temperature environments and offer enhanced wear resistance but provide lower corrosion protection, especially under acidic conditions.

Boron-based electroless nickel baths are classified into low-boron (0.5-3% B) and high-boron (3-5% B) variants, each offering distinct properties. Low-boron coatings enhance their solderability and conductivity, making them well suited for electronic applications. In contrast, high-boron coatings exhibit increased hardness and improved resilience after heat treatment, making them a viable alternative to chrome coatings in various industries, although they are less effective in corrosive environments. However, both types of coatings

tend to be more porous than the other electroless coatings, which can limit their performance in certain applications.

By adding various elements to electroless nickel coatings, new alloys (such as binary, ternary, or polyalloys) can be formulated to achieve specific properties. This flexibility allows the creation of coatings tailored to meet a wide range of industrial requirements, such as improved corrosion resistance, wear resistance, lubricity, and electrical conductivity. The ability to customize these coatings makes them highly versatile and valuable for diverse industrial applications.

Nickel composite coatings

These coatings incorporate second-phase materials such as ceramic particles (e.g., TiO_2 , Al_2O_3 , and SiC), minerals (diamond and columbite-tantalite), and polymers (polyvinyl chloride, PTFE) within the nickel matrix, significantly enhancing characteristics such as wear resistance and self-lubricating properties. Nickel composite coatings are gaining traction for applications requiring strong mechanical performance, such as in the aerospace and automotive industries, owing to their ability to withstand extreme conditions and provide additional functionalities.

1.3 Principles of electroless coating deposition

Electroless coating is a non-galvanic metal deposition method that uses a chemical reducing agent to convert metal ions to a solid metal state in an aqueous solution. The process is highly autocatalytic, allowing uniform coating deposition across various substrate geometries [37].

Key components of electroless coating:

i. Metal ion source:

This process typically uses metal salts, with nickel (Ni) being one of the most common salts, often in the form of nickel sulfate (NiSO_4). Other metal ions, such as copper (Cu) and silver (Ag), can also be plated.

ii. Reducing agent:

A chemical reducing agent is required to facilitate the transfer of the electrons necessary for the reduction of metal ions. Common reducing agents include sodium hypophosphite (NaH_2PO_2), hydrazine, or glucose.

iii. Complexing agents:

Complexing agents, such as citrates or ammonium salts, are often used to stabilize metal ions in solution and prevent premature precipitation, complexing agents (like citrates or ammonium salts) are often used. These agents help to maintain the metal ion concentration and pH within the desired range.

iv. pH buffering agents:

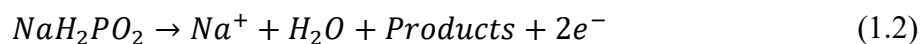
The pH of the solution affects the deposition rate and quality of the coating. Adjusting agents are used to maintain an optimal pH, usually between 4.5 and 5.5, for nickel coating. The fundamental reactions in electroless coatings involve oxidation and reduction reactions, which can be broken down as follows:

- i. Cathodic reaction (reduction): Ni ions are reduced to form a solid Ni.



Here, Ni^{2+} ions in the solution gain electrons supplied by the oxidizing agent, forming metallic Ni on the substrate.

- ii. Anodic reaction (oxidation): the reducing agent is oxidized to release electrons.



Here, sodium hypophosphite undergoes oxidation, contributing to the electrons needed to reduce nickel ions.

1.3.1 Thermodynamics of electroless coating

Thermodynamics deals with the energy changes associated with chemical reactions and, in the context of electroless coating, provides insights into the feasibility and driving forces of the deposition process. The feasibility is governed by Gibbs free energy (ΔG), where a negative ΔG indicates a spontaneous reaction. The process is also driven by the electrode potentials of the half-cell reactions, that is, the cathodic reaction ($Ni^{2+} \rightarrow Ni$) and the anodic reaction (oxidation of reducing agents). A positive overall cell potential implies favorable deposition.

Mixed Potential Theory explains that the actual deposition potential (mixed potential) results from simultaneous anodic and cathodic reactions. This steady-state potential is vital for understanding and controlling the spontaneity and balance of a reaction.

1.3.2 Kinetics of electroless coating

Kinetics studies the rate of metal deposition influenced by temperature, reactant concentration, nature of the reducing agent, and surface conditions. The rate law is often as follows:

$$R = k[M^{2+}][Red]^n \quad (1.3)$$

where R is the deposition rate, k is the rate constant, $[M^{2+}]$ is the concentration of metal ions, and [Red] is the concentration of the reducing agent.

The temperature dependence of the deposition rate can be understood using the Arrhenius equation:

$$k = Ae^{-\frac{E_a}{RT}} \quad (1.4)$$

where k is the rate constant, A is the pre-exponential factor, E_a is the activation energy, R is the gas constant, and T is temperature (K). A lower activation energy indicates a higher reaction rate.

Mass transport (diffusion) and reaction control play key roles in this process. In diffusion-limited cases, agitation improves the rate. In kinetically controlled systems, the surface reactions are dominant.

Hydrogen evolution can occur as a side reaction, particularly at a certain pH or with specific reducing agents, affecting the coating quality. Managing this is essential for the deposit efficiency and selectivity.

1.3.3 Role of catalysts and surface activation

The substrate surface must be sufficiently activated to facilitate an electroless coating. Activation may involve:

- Surface preparation: Proper cleaning and conditioning of the surface ensure that it is free from contaminants and can bond with the deposited metal.

- Catalytic activation: Palladium and other noble metals are often used as catalysts to provide the initial sites for nickel deposition. The process begins in the activated regions, allowing Ni deposition to proceed in an autocatalytic manner, where the deposit enhances further deposition.

1.3.4 Parameters influencing electroless coating

i. pH control:

The solution pH plays a significant role in controlling the deposition rate and morphology of the deposited film. Lower or higher pH levels can lead to poor deposit quality or the formation of unwanted by-products.

ii. Temperature:

The reaction rate increases with temperature owing to the enhanced diffusion and kinetic energy of the reactants. However, excessively high temperatures can lead to rapid decomposition of the bath or uneven deposition.

iii. Concentration of components:

The concentrations of metal ions, reducing agents, and stabilizers can significantly affect the rate and uniformity of the deposition. Therefore, optimal concentrations are necessary to obtain high-quality deposits with desirable properties.

iv. Time of deposition:

The deposition time affects the thickness and quality of the coating. Controlling the deposition time is necessary to prevent excessive build-up or weak adhesion.

1.4 Engineering multifunctional properties in electroless coatings

1.4.1 Hardness and wear resistance

Electroless coatings reinforced with particles, such as Al_2O_3 , SiC , and ZrO_2 , exhibit significantly improved hardness and tribological performance. Al_2O_3 is particularly effective owing to its low friction, high wear resistance, and chemical stability when incorporated into Ni-P matrices [38–40]. Refractory metals such as Mo and W further enhance thermal stability and toughness [41], while MWCNTs contribute to self-lubricating properties that reduce friction and wear [42].

The hardness and wear resistance of the composite coatings were strongly influenced by the dispersion and content of reinforcing particles. A higher nanoparticle content in Ni-ZrO₂ and Ni-Al₂O₃ coatings increases the hardness and decreases the wear rates [40]. Similarly, Cu-coated B₄C particles enhance the mechanical strength and wettability of Al2025 alloys [43], and Ni-P-Cu coatings benefit from phase formation (Ni₃P, NiCu), which improves tribo-mechanical behavior through grain refinement and dislocation inhibition [44].

Post-deposition heat treatment at 400 °C promoted the crystallization of Ni-P to Ni and Ni₃P phases, increasing the hardness and wear resistance [45]. However, temperatures above 450 °C can degrade performance owing to oxidation, grain coarsening, and Fe diffusion, leading to reduced adhesion and increased porosity [44].

1.4.2 Corrosion resistance

Electroless composite coatings are known to enhance corrosion resistance. Ni-P/PTFE coatings reduce the corrosion current and increase the corrosion potential, although they may compromise hardness and stiffness [46]. Carbon nanoparticles, such as those from candle soot, improve corrosion resistance cost-effectively in Ni-P coatings [47]. Bi-layer coatings, such as Co/Sn, offer multilayer protection by reducing the corrosion rates [48].

Alloying Ni-P/Ni-B with elements such as W and La³⁺ enhances the corrosion behavior through oxide film formation and grain refinement, although optimal concentrations are critical to avoid performance decline [49,50]. Ni-P-Cu coatings benefit from optimized copper levels and heat treatment, which improve the morphology and protective properties [45].

Ni-Mo-P coatings provide high electrochemical stability in aggressive environments, making them suitable for energy and marine applications [51]. Coatings produced using wastewater also show good corrosion resistance at moderate temperatures, although their performance drops beyond 400 °C owing to oxide-related degradation [52].

1.4.3 Thermal stability

Electroless coatings play a significant role in enhancing the thermal stabilities of various composite materials. The deposition of Ag and Ni onto h-BN and CNTs improves the interfacial bonding and dispersion within the aluminum matrix composites, reducing

agglomeration and oxidation, and improving the high-temperature performance [53]. Similarly, cerium-containing coatings increase the oxidation resistance by forming dense oxide scales and acting as cation diffusion barriers [54].

Ag-coated glass flakes improve the thermal and mechanical performance of PAE composites by increasing their degradation and softening temperatures [55]. In thermoelectric materials, electroless Ag/SnTe coatings enhance the Seebeck coefficient, reduce the thermal conductivity, and improve the overall thermoelectric efficiency ($zT = 0.67$ at 823 K) [56].

Thermal stability is also improved in Ag nanowire-based electrodes through AgNP deposition, significantly lowering the sheet resistance and preserving the conductivity under heat exposure [57]. Ni-Fe-P coatings on bamboo fibers increase the degradation temperature and reduce pyrolysis, thereby acting as a protective thermal barrier [58]. Additionally, ScF₃ electroless coatings enable near-zero thermal expansion in Cu composites, providing thermal shock resistance ideal for electronic packaging applications [59].

1.4.4 Antimicrobial properties for biomedical and marine applications

Electroless nanocomposite coatings have shown promising antimicrobial performances, making them suitable for applications in biomedical and marine environments. NiP-TiNi coatings reduced *E. coli* viability by over 80% and inhibited biofilm formation, achieving up to 93% corrosion inhibition in sulfate-reducing bacterial media [60]. Similarly, Ni-P coatings enhanced with ZnO-doped C₃N₄ nanocapsules effectively reduced *S. aureus* growth, particularly at low ZnO concentrations, with heat treatment further improving bactericidal activity [61].

Silver-based coatings also demonstrate strong antimicrobial effects. Ag-loaded ceramic papers and Ag-coated PA6 fibers showed over 99% bacterial inhibition against *S. aureus* and *E. coli*, with durability retained after washing, indicating potential in food preservation and textile applications [62,63]. Electroless Ni-P coatings with ZnO nanoparticles on ABS substrates further reduce bacterial colonization, reinforcing their relevance for medical devices and hygiene-critical surfaces [64].

Cerium- and neodymium-containing coatings also contribute to antimicrobial activity. Cerium compounds disrupt bacterial structures and generate reactive oxygen species,

thereby offering dual corrosion and microbial resistance [65]. Nd₂O₃-Ag nanohybrids exhibit over 98% bactericidal efficiency, combining ion release and membrane disruption mechanisms [66].

1.5 Optimization of electroless coating process parameters

1.5.1 Traditional optimization techniques

Early attempts to optimize the electroless coating processes have been largely empirical. Researchers have systematically varied individual parameters while observing their effects on the resulting coating properties. This "one-factor-at-a-time" approach, while conceptually simple, proved to be highly inefficient and time-consuming. Furthermore, it fails to account for the potential interactions between multiple parameters, leading to suboptimal results. The inherent limitations of this method necessitate a shift towards more statistically rigorous approaches [67,68].

The development of the statistical design of experiments (DoE) has marked a crucial turning point. Techniques, such as orthogonal arrays [69] and the Taguchi method [70], provide frameworks for systematically varying multiple parameters simultaneously, enabling a more efficient exploration of the parameter space. In particular, the Taguchi method has gained significant traction owing to its ability to minimize the number of experiments required, while still providing statistically robust results. By employing orthogonal arrays, researchers can identify significant factors and their optimal levels with fewer experimental runs than traditional full-factorial designs. Studies using the Taguchi method have successfully optimized various aspects of electroless coatings, including the microhardness and deposition rate [71,72]. However, even these advanced DoE methods often struggle to adequately capture the complex interactions between multiple parameters, potentially leading to incomplete or misleading conclusions [73]. This limitation has spurred the development of more sophisticated modelling techniques.

1.5.2 Response surface methodology (RSM)

Response surface methodology (RSM) has emerged as a powerful statistical tool for modelling and optimizing complex processes characterized by multiple interacting variables. RSM utilizes mathematical and statistical techniques to construct response surfaces, which graphically represent the relationship between the process parameters

(independent variables) and the desired coating properties (dependent variables, such as corrosion rate, hardness, or deposition rate) [74,75]. RSM employs various experimental designs, including Box-Behnken [76] and central composite designs [75], to gather data for model construction. These designs were chosen for their efficiency in exploring the parameter space, while minimizing the number of required experiments. The resulting empirical models, often second-order polynomial equations, allow for the prediction of optimal parameter combinations that yield the desired coating characteristics [77].

The CCD structure for the three-factor design is shown in Figure 1.5. It comprises three components: (1) a factorial or fractional factorial design with combinations of factors at two levels (high and low), represented by corner points; (2) center points to estimate experimental error and curvature; and (3) axial (star) points placed at a distance (α) from the center to ensure rotatability, providing uniform precision in all directions.

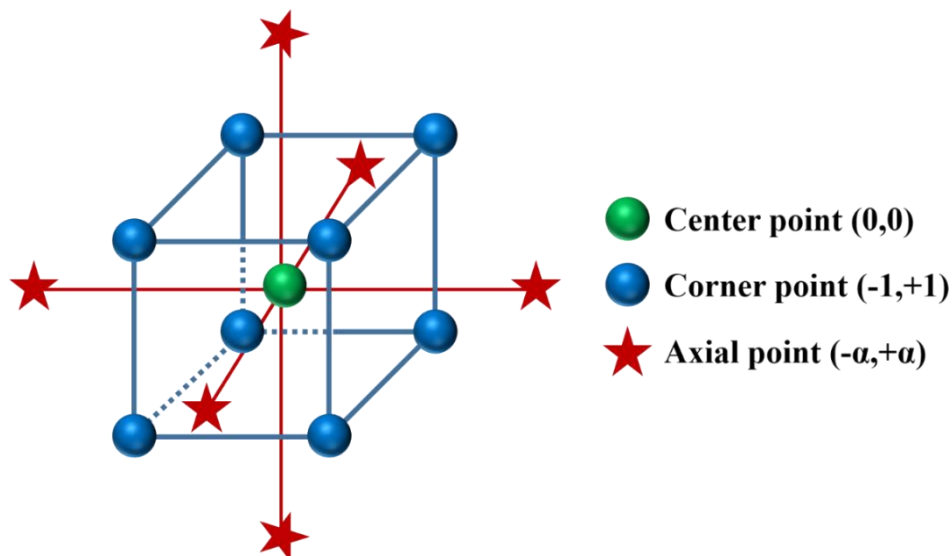


Figure 1.5. Structure of CCD for the three-factor experimental design.

1.5.3 Artificial neural networks (ANNs)

Artificial neural networks (ANNs), inspired by biological neural networks of the brain, provide a powerful machine-learning approach for modelling complex nonlinear systems. Unlike the RSM, which relies on explicitly defined mathematical functions, ANNs can learn complex relationships from data without requiring a priori assumptions regarding the underlying functional form. This makes them well suited for modelling electroless coating processes, where the interplay between parameters can be highly intricate and nonlinear [78].

1.5.4 Metaheuristic algorithms

Metaheuristic algorithms offer a powerful class of optimization methods inspired by natural processes, such as swarm intelligence, biological evolution, and physical phenomena. These algorithms are particularly well-suited for tackling complex multimodal optimization problems, where traditional gradient-based methods may fail to find the global optimum [79]. Their robustness and ability to handle noisy or incomplete data make them attractive for optimizing the electroless coating processes. Several metaheuristic algorithms have been applied to this field, including

Genetic algorithm (GA): Inspired by the principles of natural selection and genetic evolution, GAs have been widely employed for optimization problems across various engineering disciplines [70,77,80]. Their ability to handle complex, multi-objective optimization problems and their robustness to noise make them particularly well-suited for optimizing the electroless coating process. GAs operate by evolving a population of potential solutions and selecting the fittest individuals based on a fitness function that evaluates the quality of the solutions. Through genetic operators such as crossover and mutation, new solutions are generated, and the population evolves towards better solutions over generations [81].

Several other swarm intelligence-based metaheuristic algorithms have also been applied to optimize electroless coating parameters, including the Artificial bee colony (ABC) algorithm, Firefly algorithm (FA), and Teaching-learning-based optimization (TLBO). The ABC algorithm models the foraging behavior of honey bees by utilizing employed bees, onlookers, and scout bees to efficiently explore and exploit the search space for optimal solutions. The Firefly Algorithm is inspired by the mutual attraction among fireflies, where individuals are drawn toward brighter, more attractive solutions, directing the search for optimal outcomes. Meanwhile, TLBO replicates a classroom learning process in which learners enhance their knowledge through interaction with a 'teacher,' representing the best solution and their peers, and progressively improving solution quality through iterative learning phases [76].

1.5.5 Hybrid approaches

The inherent strengths of different optimization techniques have motivated the development of hybrid approaches that combine the capabilities of multiple methods. These

hybrid models aim to leverage the advantages of each component technique, while mitigating their individual limitations. For example, integrating ANNs with metaheuristic algorithms combines the powerful predictive capabilities of ANNs with the global search capabilities of metaheuristics [81,82]. This synergistic approach can lead to a more accurate and efficient optimization of the electroless coating parameters.

Similarly, combining RSM with metaheuristic algorithms can enhance the exploration of the parameter space and improve the robustness of the optimization process [83]. RSM provides a statistical framework for modelling the relationships between process parameters and coating properties, whereas metaheuristics offer a powerful mechanism for searching for optimal solutions in the parameter space defined by the RSM model. The combination of these two approaches can lead to more reliable and accurate predictions of the optimal parameter combinations. Other hybrid methods include gray-fuzzy-based approaches [69] that combine statistical modelling with fuzzy logic to handle uncertainty and vagueness in the optimization process.

1.6 Research motivation and scope

1.6.1 Development of electroless polyalloy coatings on low-carbon steel

Electroless polyalloy coatings, particularly Ni-P-based alloys, offer a promising solution for enhancing the properties of low-carbon steel by incorporating elements such as Mo, W, and Co. These coatings improve hardness, corrosion and wear resistance, thermal stability, frictional properties, and antimicrobial effectiveness, making them suitable for multifunctional applications. Given the widespread industrial use of low-carbon steel and its inherent limitations in corrosion resistance and wear durability, advanced coatings are essential for extending the component lifespan and improving performance in demanding environments. This study aims to develop a novel Ni-P alloy coating tailored for multifunctional applications in marine, biomedical, antifouling, and other industrial sectors.

1.7 Thesis outline and contributions

The thesis entitled “Development, characterization and optimization of electroless Ni-W-P-Ce coatings for multifunctional applications” has been divided into seven chapters, excluding the appendix. An overview of the remainder of this thesis is summarized below.

Chapter 2 presents a literature review focusing on the development of multifunctional electroless polyalloy coatings. It summarizes previous research on polyalloy coatings for various applications and highlights the key findings and advancements. Through this review, existing research gaps were identified, which formed the basis for defining the objectives of this study.

Following that, **Chapter 3** describes the synthesis and characterization of rare earth metal cerium-added electroless Ni-W-P coatings on low-carbon steel. This study examines the effects of varying the chemical concentrations and coating parameters to achieve a stable coating bath and high-quality Ni-W-P-Ce coated samples.

Afterward, in **Chapter 4**, 30 experiments were designed using the central composite design within the response surface methodology, with the lower and upper factor levels derived from the findings of the previous chapter. Subsequently, multi-objective optimization of the electroless Ni-W-P-Ce coating parameters was conducted using a hybrid ANN-metaheuristic optimization technique to simultaneously enhance five key coating properties: roughness, hardness, antifouling ability, surface free energy, and corrosion protection efficiency.

Subsequently, **Chapter 5** explores the microstructural evolution and performance characteristics of cerium-modified electroless Ni-W-P coatings, both before and after heat treatment. The coatings were deposited using optimized parameters, and only the Ce content was varied. This study systematically examined the effect of the cerium concentration on the deposition rate, microstructure, composition, and functional properties.

In continuation, **Chapter 6** examines the influence of key process parameters- activator solution, bath loading, and substrate roughness- on the initiation, growth, and properties of electroless Ni-W-P-Ce coatings deposited on low-carbon steel substrates.

Finally, **Chapter 7** presents a summary of the research work, along with recommendations for future research directions. The references used in this study are listed at the end of this thesis.

Chapter 2 Literature review

This chapter reviews electroless coatings with a focus on their composition, multifunctional properties, and enhancement strategies, including the use of rare earth elements. It also examines optimization techniques, such as statistical methods, machine learning, and metaheuristic algorithms, and concludes with identified research gaps and objectives that frame the study.

2.1 Role of alloying elements in electroless coatings

The incorporation of secondary elements into electroless Ni-P coatings is a proven strategy for enhancing their properties and expanding their applications. Elements such as tungsten (W), zinc (Zn), copper (Cu), chromium (Cr), molybdenum (Mo), and rare earth elements significantly improve mechanical and physical characteristics, including hardness, wear resistance, corrosion resistance, thermal stability, and magnetic properties, depending on their concentration, bath composition, and deposition parameters. Multi-element alloying further enhances the coating performance, enabling customized functionalities for specific applications. For instance, hard particles or lubricating agents improve wear resistance and reduce friction, whereas noble metals enhance corrosion protection in aggressive environments [84,85].

The specific effects of individual elements on the microstructure and performance of Ni-P based polyalloy coatings are discussed in the following subsections.

2.1.1 Role of cobalt

The incorporation of cobalt into electroless Ni-P coatings resulted in 12.67 wt.% cobalt and 14.63 wt.% phosphorus. Co significantly contributed to the magnetic properties of the Ni-Co-P coatings. In a quaternary system, where both cobalt and cerium were introduced into the Ni-P bath, the annealed Ni-Co-P-Ce coating (at 400 °C) exhibited a composition of 21.5 wt.% cobalt, 7.1 wt.% phosphorus, and 1.3 wt.% cerium. The presence of cobalt enhanced nickel ion deposition, facilitated the formation of the ferromagnetic Ni(Co) phase, and suppressed the development of paramagnetic phases, such as Ni₃P and Co₂P. Consequently, the coating demonstrated improved saturation magnetization and coercivity [84,86].

2.1.2 Role of iron

Iron (Fe) plays a pivotal role in shaping the structural and functional characteristics of electroless Ni-Fe-P coatings. The Fe and phosphorus (P) contents of these coatings can vary significantly, with Fe ranging from 3 wt.% to 24 wt.% and P from 4 wt.% to 14 wt.%. The amount of iron directly influences the microstructure; coatings with less than 9 wt.% Fe are typically amorphous; those with Fe content between 9 wt.% and ~12 wt.% exhibit a mixed amorphous-crystalline structure; and above 12 wt.%, the coatings become fully crystalline. This structural evolution contributes to enhanced thermal stability and improved wettability, making Ni-Fe-P coatings suitable candidates for applications such as under-bump metallization (UBM) in high-temperature electronic packaging [87].

In more complex systems, such as quaternary Ni-Co-Fe-P coatings, the introduction of iron exerts a synergistic effect with nickel and cobalt, significantly enhancing the electrocatalytic performance. Iron promotes the formation of a mixed amorphous and nanocrystalline structure, which increases the density of electrochemically active sites and is especially relevant for the hydrogen evolution reaction (HER). Moreover, an optimized atomic ratio of Ni, Co, Fe, and P has been found to correlate with improved HER activity, highlighting the critical role of Fe in tuning the catalyst composition for maximum performance [88].

2.1.3 Role of copper

Copper (Cu) incorporation into electroless Ni-P coatings has been shown to markedly enhance their functional properties, particularly in Ni-P-Cu composite systems. The addition of copper improves corrosion resistance and heat transfer performance owing to its high thermal conductivity. It also refines the microstructure, resulting in a denser coating with an increased mechanical strength. Studies have demonstrated that coatings with an optimal Cu concentration (1 g/L) significantly outperform uncoated mild steel, with cooling rates improved by approximately 25.2%, and notable increases in corrosion resistance were observed within the 100-400 °C range. These enhancements make Ni-P-Cu coatings highly suitable for thermal management applications, such as heat exchangers [45].

In quaternary systems such as Ni-P-Cu-W coatings, copper contributes to modifying the surface texture and grain structure, leading to larger grain sizes and improved mechanical characteristics. After annealing at 500 °C, the coatings exhibited an elastic

modulus of approximately 10.48 GPa. Despite earlier reports suggesting the limited influence of copper on tribological properties compared to tungsten, the presence of copper in these polyalloy coatings resulted in a significant reduction in the wear rate and coefficient of friction. This indicates a synergistic effect, where copper aids in enhancing the tribological performance when combined with tungsten, thus broadening the functional applications of these coatings [89].

2.1.4 Role of zinc

Zinc (Zn) plays a vital role in enhancing the performance of electroless Ni-P-based coatings. In the Ni-Zn-P system, elemental analysis revealed weight percentages of 7.24% Zn, 9.4% P, and 83.36% Ni. The incorporation of Zn significantly improves the mechanical strength and corrosion resistance of the coatings. Notably, Zn doping has been shown to enhance hardness, fatigue life, and electrochemical behavior, making Ni-Zn-P coatings suitable for applications requiring durable and protective surfaces [90].

Zn offers substantial benefits to quaternary Ni-Zn-Cu-P coatings. The elemental composition of the duplex coating was Ni (79.71 wt.%), P (10.25 wt.%), Zn (7.55 wt.%), and Cu (2.5 wt.%). The introduction of Zn in this system contributed to the development of a more compact, uniform granular microstructure, which significantly improved the corrosion resistance compared to conventional Ni-P coatings, particularly under 3.5 wt.% NaCl solution testing. Moreover, while Zn^{2+} ions are known to inhibit plating rates, the addition of Cu^{2+} ions effectively balances this effect, resulting in coating thicknesses comparable to those of binary systems. This synergy highlights the multifaceted role of Zn in improving both the structural integrity and corrosion performance of electroless alloy coatings [91].

2.1.5 Role of molybdenum

Molybdenum (Mo) has a significant influence on the corrosion resistance and mechanical performance of electroless Ni-P coatings. Its incorporation notably improved the resistance to pitting corrosion, particularly in chloride-rich environments, by promoting the formation of a stable passive film on the coating surface. Furthermore, Mo enhances the adhesion of the coating to the magnesium alloy substrates. Optimal results were observed at a Na_2MoO_4 concentration of 0.4 g/L, which provided a balanced Mo and phosphorus (P) content, resulting in improved microstructure and mechanical strength [92].

The presence of Mo further enhanced the thermal stability and structural integrity of Ni-Mo-W-P quaternary coatings. Specifically, Mo helps stabilize the amorphous phase up to 450 °C, effectively delaying the phase transformations that typically result in brittleness. This stabilization preserves the fracture toughness and allows the alloy to achieve higher hardness levels without compromising the ductility. These characteristics make Ni-Mo-W-P coatings especially advantageous for high-temperature applications, including solderable surfaces in integrated circuits, where both durability and thermal resilience are essential [41].

2.1.6 Role of tungsten

Tungsten (W) is a critical alloying element in electroless Ni-P coatings and contributes significantly to corrosion resistance and structural enhancement. Its inclusion in Ni-W-P coatings promotes the formation of a robust passive film that effectively inhibits the ingress of water and oxygen, thereby providing excellent protection to the underlying magnesium alloy substrates. Additionally, tungsten plays a role in refining the grain structure of the coating and increasing the nucleation rate during deposition, which helps to maintain an amorphous structure. These effects result in a more uniform, dense, and protective coating surface with improved durability [93].

Tungsten has been shown to considerably improve the electrocatalytic performance in Ni-W-Mo-Co-P polyalloy coatings. The uniform distribution of W ions throughout the coating enhances its intrinsic electrochemical characteristics and alters its electronic structure, leading to an increased active surface area. This contributes to the superior HER activity, with reported overpotentials of 41 mV at 10 mA/cm², 46 mV at 20 mA/cm², and 67 mV at 100 mA/cm², along with a favorable Tafel slope of 38 mV·dec⁻¹. The electroless deposition technique further ensures the chemical stability and durability of the Ni-W-Mo-Co-P coatings, making them excellent candidates for energy-related applications that require robust and efficient HER catalysts [94].

2.1.7 Role of rare earth elements

The incorporation of rare earth elements (REEs) such as lanthanum (La), yttrium (Y), cerium (Ce), and neodymium (Nd) significantly enhances the microstructure, deposition rate, and overall performance of the electroless coatings. REEs improve bath stability, facilitate catalytic reduction, and refine grain structures, leading to denser coatings with

superior corrosion resistance and mechanical properties [50,95]. Optimized concentrations of REEs accelerate deposition rates by increasing the number of nucleation sites, whereas excessive amounts can hinder the process by shielding the catalytically active sites [95]. In Ni-P-B coatings, REEs, such as La^{3+} and Ce^{3+} , promote finer grain structures and self-cleaning superhydrophobic surfaces, enhancing their protective capabilities [50]. Similarly, the addition of Ce and Nd in Ni-W-P and Ni-Mo-P coatings results in improved adhesion, reduced surface defects, and lower corrosion rates, making them ideal for applications in aerospace, automotive, and other industries exposed to harsh environments [96,97]. REEs also contribute to the amorphization of coatings, enhancing their durability and electrochemical stability [96]. In addition to corrosion resistance, REEs improve catalytic activity in the hydrogen evolution reaction (HER), as seen in the Ni-Co-B-RE coatings. Elements such as gadolinium (Gd) optimize the electronic structure and enhance electron transfer, leading to a lower overpotential and improved hydrogen adsorption/desorption kinetics [98]. Furthermore, the incorporation of Ce in Ni-Cu coatings modifies the reduction potential of copper, forming a denser passive film that reduces pitting corrosion and extends the service life of coatings in corrosive environments [99]. Overall, the strategic incorporation of REEs into electroless coatings significantly enhances their structural integrity, tribological properties, and corrosion resistance, rendering them valuable for advanced engineering applications [100].

Role of cerium and its potential benefits in electroless coatings

Cerium (Ce) has been used for the development of advanced functional coatings owing to its exceptional physicochemical properties, including corrosion resistance, self-healing capabilities, superhydrophobicity, and UV shielding. The most common applications of Ce are shown in Figure 2.1. As one of the most abundant rare-earth elements, the dual-valence transformation of cerium (Ce^{3+} and Ce^{4+}) enhances its effectiveness as a corrosion inhibitor, forming protective passive layers on metal surfaces that significantly reduce corrosion rates. The integration of cerium compounds into various coating matrices not only improves their durability and performance but also opens avenues for innovative applications in diverse fields such as aerospace, automotive, and electronics. By fully utilizing the unique characteristics of Ce, researchers aim to address the challenges of resource waste and expand the functional applications of rare-earth elements in coatings, ultimately contributing to more sustainable manufacturing practices and increasing economic value [65].

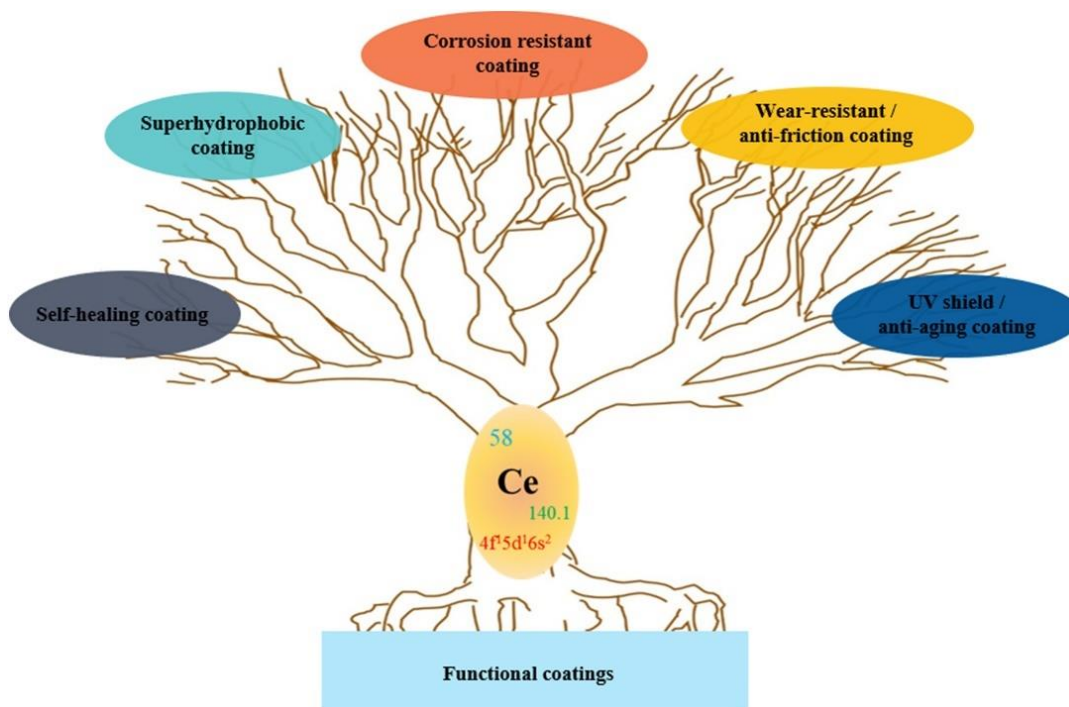


Figure 2.1. Applications of rare earth metal cerium (adapted from [65]).

In electroless coatings, such as Ni-P, the incorporation of cerium oxide (CeO_2) not only promotes structural stability but also improves the electrochemical properties. The presence of Ce enhances charge-transfer kinetics through the formation of oxygen vacancies, which facilitates electron transfer. Furthermore, the defective CeO_2 species contributed to the regulation of the adsorption-desorption processes, thereby improving the overall hydrogen evolution efficiency [101].

Microstructural changes

The incorporation of Ce into the electroless coatings can induce significant microstructural changes. Cerium ions can act as nucleation sites, influencing the growth of the coating and leading to the refinement of the grain size. This grain refinement can enhance the mechanical properties of the coating, such as hardness and wear resistance. The presence of Ce can also affect phase formation in the coating, leading to the formation of different intermetallic compounds or oxide phases [102]. These microstructural changes can significantly affect the overall performance of the coating, thereby affecting its corrosion resistance, thermal stability, and other properties [97]. The specific microstructural changes observed depended on various factors, including the type of coating, cerium concentration, and deposition conditions.

Mechanical properties

The addition of cerium to electroless coatings often leads to improvements in mechanical properties, particularly hardness and wear resistance. This enhancement can be attributed to several factors, including grain refinement, dispersion strengthening [103], and formation of hard intermetallic phases [102]. Studies have shown that optimal Ce concentrations can significantly increase the microhardness of Ni-P [102], Ni-B, and other coatings [104]. This improved hardness translates to enhanced wear resistance, as demonstrated by various tribological tests. However, exceeding the optimum cerium concentration can lead to a decrease in wear resistance, indicating the importance of precise control over cerium content. The underlying mechanisms responsible for the improved wear resistance include increased grain boundary strength and formation of protective oxide layers [105].

Corrosion resistance

The incorporation of Ce into electroless coatings significantly enhanced their corrosion resistance. This improvement could be attributed to several factors. First, cerium forms a passive oxide layer on the coating surface, acting as a barrier against corrosive agents [106]. Second, Ce ions can block the diffusion paths of the corrosive species, preventing them from reaching the substrate. Third, Ce can react with aggressive anions (e.g., chlorides) to form insoluble compounds, further enhancing corrosion protection [54]. Studies have shown that the corrosion resistance of various coatings, including Ni-Ce-P [54], Ni-Mo-P [97], and Ni-B-CeO₂ [105], is significantly improved by the addition of Ce. The optimal cerium concentration for achieving maximum corrosion resistance varies depending on the specific coating system and the corrosive environment. The effectiveness of Ce in enhancing corrosion protection is evident in various environments including saline solutions [97], molten salts [106], and aggressive atmospheric conditions [107]. The formation of protective cerium oxide layers [108] and blocking of diffusion paths contributed to the enhanced corrosion resistance observed in these studies. The type of phosphate post-treatment also plays a crucial role in determining the corrosion resistance of cerium-containing coatings, with different phosphate treatments leading to significant variations in stability against pitting corrosion [107].

Thermal stability

The addition of Ce significantly improves the thermal stability of the electroless coatings. This enhancement was attributed to the ability of Ce to inhibit the diffusion of phosphorus [109] or boron [85] atoms within the coating matrix. These atoms are often responsible for the degradation of the coating at elevated temperatures. The presence of Ce hinders the mobility, thus increasing the temperature at which significant structural changes occur. This improved thermal stability is crucial for applications involving high-temperature exposure, such as thermal barrier coatings [85]. The specific mechanisms of the thermal stability enhancement vary depending on the coating composition and cerium concentration.

Other properties (e.g., magnetic, biofilm inhibition)

Cerium incorporation can also influence other properties of electroless coatings depending on the specific composition and processing parameters. For instance, the addition of Ce to Ni-Co-P coatings enhances their magnetic properties, leading to improved saturation magnetization. The underlying mechanism involves the promotion of ferromagnetic phase formation in the coating [110]. In other cases, Ce has been found to improve the biofilm inhibition capabilities of coatings, making them suitable for applications where microbiologically induced corrosion is a concern [102]. These findings highlight the versatility of Ce as an additive in electroless coatings and its potential to tailor the properties of the coatings to meet specific application requirements.

2.2 Electroless coating process parameter optimization**2.2.1 Process optimization via response surface methodology**

Response surface methodology (RSM), often combined with design of experiments (DoE) techniques such as central composite design (CCD) and Box-Behnken design, have been extensively employed to optimize the deposition parameters and enhance the performance of electroless coatings.

Zhang et al. [74] utilized RSM with CCD to optimize the parameters for Ni-Fe-P electroless coatings. A full factorial design involving ten experimental runs was conducted, followed by analysis of variance (ANOVA) to validate the predictive models. The optimal coating performance was achieved at a bath temperature of 85.1 °C and a pH of 7.9, leading

to a significant enhancement in corrosion resistance. The optimized coating exhibited a polarization resistance of $22,180.8 \Omega \cdot \text{cm}^2$, corrosion current density of $3.30 \mu\text{A}/\text{cm}^2$, and corrosion potential of -0.4253 V .

Similarly, CCD within the RSM framework was applied to investigate the effects of cobalt sulfate, sodium hypophosphite, and bath temperature on the corrosion behavior of Ni-Co-P coatings. The optimal conditions; 15 g/L cobalt sulfate, 30 g/L sodium hypophosphite, and 80°C bath temperature, yielded a minimum corrosion rate of $0.535 \mu\text{m}/\text{year}$, representing a 74.76% reduction from the original copper substrate rate of $2.12 \mu\text{m}/\text{year}$ [75].

Further optimization of the Ni-Co-P coatings was explored using the Box-Behnken design across three variables: nickel sulfate, cobalt sulfate, and bath temperature. Fifteen experimental runs were conducted, and the data were analyzed using machine learning algorithms, such as gradient descent with RMSprop, ABC, Firefly, and Teaching-learning-based optimization. The study identified an optimal corrosion rate of $0.435 \mu\text{m}/\text{year}$, with statistical analysis confirming the strong influence of the cobalt sulfate concentration and its interaction with bath temperature on the coating performance [76].

In the case of electroless Ni-W-P coatings applied to AZ91D magnesium alloys, RSM was employed to develop a quadratic regression model assessing multiple process parameters. Optimal conditions included 20 g/L nickel sulfate, 15 g/L sodium tungstate, 30 g/L sodium hypophosphite, a bath temperature of 65°C , and a pH of 9.3. Among these, the sodium tungstate concentration was the most influential, with interactions among variables that significantly affected the corrosion resistance of the coating. The optimized coating demonstrated an impressive failure time of 10,200 min in immersion tests, indicating strong protective capabilities [111].

Optimization of the Ni-B-nanodiamond composite coatings was carried out using RSM with a CCD framework involving 15 experiments. The variables included the nanodiamond concentration, reducing agent concentration, and stabilizer levels. The resulting quadratic models accounted for 94.85% of the response variability in the microhardness and thickness. Under optimal conditions, the coatings achieved a maximum microhardness of 871 HK50 and a thickness of up to $15.0 \mu\text{m}$, confirming the effectiveness of RSM in tailoring the coating properties [112].

2.2.2 Modeling process-property relationships using artificial neural networks

Vijayanand et al. [113] applied a back-propagation neural network (BPNN) to optimize the hardness of citrate-stabilized electroless nickel-boron coatings by modeling the effects of nickel concentration, reducing agent, and stabilizer. The model was trained using data from 11 experiments and validated using four additional runs. It achieved a high coefficient of determination ($R^2 = 0.9852$), with a predicted maximum hardness of 592 HV, which closely matched the experimental results. Although the ANN model was highly effective, the Box-Behnken design (BBD) offered a slightly better fit, achieving $R^2 = 0.9911$.

Vijayanand et al. [114] in another study used BPNN to explore the relationship between the concentration of an amphoteric surfactant and the average surface roughness (R_a) of electroless Ni-B coatings. Using a multilayer perceptron structure with optimized parameters, the network achieved $R^2 = 0.988$, outperforming traditional regression approaches and effectively capturing the influence of the surfactant concentration on the surface morphology.

In a different dipping and padding coating process, a cascade-forward neural network was used to predict the concentration of silver nanoparticles (AgNPs) coated on dyed silk fabric using bacterial inhibition zones as the input. The network architecture includes multiple hidden layers with neurons connected from all the previous layers to enhance learning. This approach significantly improved the accuracy of AgNP concentration prediction compared with traditional spectrophotometric color matching, achieving a high correlation of 0.994 between the actual and predicted values and a low mean absolute error of 0.46 [115].

ANN modeling was also employed to predict the deposition rate of the Ni-W-P-nanoTiO₂ composite coatings based on parameters optimized using the Taguchi method. The most effective architecture was ANN 4-7-1, which comprised four input neurons, seven hidden neurons, and one output neuron. This configuration achieved a correlation coefficient of $R^2 = 0.9820$, demonstrating strong agreement between the predicted and experimental values and reinforcing the ANN's potential for optimizing electroless deposition parameters [71].

Similarly, a feed-forward BPNN was used to predict the microhardness of the Ni-Zn-Cu-P coatings. After evaluating various neuron configurations, the ANN 3-7-1 model was found to be the most accurate, yielding an RMSE of 8.8475 and R^2 of 0.982. The ANN

model outperformed the predictions obtained using the Taguchi method, highlighting its superior predictive capability for mechanical property optimization [78].

Additionally, an enhanced dendritic neural (DN) network optimized using the artificial hummingbird algorithm (AHA) was developed to model the wear behavior of Cu-Al₂O₃ nanocomposites reinforced with silver. This advanced model accurately predicted both the wear rate and the coefficient of friction. The results revealed that increasing the Al₂O₃ content reduced wear and improved frictional properties, and the DN-AHA hybrid outperformed conventional methods in terms of predictive accuracy [82].

2.2.3 Optimization through metaheuristic algorithms

Hassan et al. [76] utilized metaheuristic optimization algorithms, including ABC, Firefly algorithm, and Teaching-learning-based optimization, to optimize the corrosion rate of electroless Ni-Co-P alloy coatings on copper substrates. The algorithms targeted significant factors, such as cobalt sulfate concentration and bath temperature, resulting in an optimized corrosion rate of 0.435 $\mu\text{m}/\text{Y}$, which is a 79.5% reduction from the corrosion rate of pure copper (2.12 $\mu\text{m}/\text{Y}$). The optimization results were consistent with the experimental data, demonstrating the effectiveness of the algorithms in enhancing the corrosion resistance of the coatings.

Mandal et al. [70] employed several metaheuristic optimization algorithms, including GA, ABC, and ANN with Differential evolution, to optimize the microhardness of electroless Ni-Sn-P coatings. The GA achieved an optimal microhardness of 387.4328 HV, whereas the ANN predicted 376.57 HV. The experimental microhardness was confirmed at 387.43 HV, highlighting the effectiveness of the algorithms. The analysis showed that nickel sulfate and sodium hypophosphite significantly influenced the microhardness, which was supported by surface morphology evaluations using techniques such as SEM and EDX.

In one study, metaheuristic optimization algorithms, including the Plackett-Burman design for screening, Box-Behnken design for local optimization, and GA and simulated annealing (SA) for global optimization, were utilized to refine the process parameters for electroless nickel-boron coatings. The combination of these techniques led to a significant reduction in the average surface roughness, reaching a value of $0.252 \pm 0.004 \mu\text{m}$ [77].

De et al. [116] used the Non-dominated sorting genetic algorithm II (NSGA-II) for multi-objective optimization to maximize the weight percentages of Nickel and Cobalt in

electroless Ni-Co-P coatings while balancing their conflicting nature. By optimizing process parameters, such as cobalt concentration, reducing agent concentration, and bath temperature, NSGA-II produced a Pareto optimal front, presenting several optimal parameter combinations. The results demonstrated a strong correlation between the NSGA-II-predicted values and actual experimental data, thus validating the approach and enhancing the efficiency of the coating process.

2.2.4 Hybrid optimization strategies

Agrawal and Mukhopadhyay [81] integrated an ANN with a GA to identify the optimal parameters for minimizing the wear rate and coefficient of friction. The ANN, trained using the Levenberg-Marquardt algorithm, predicted performance outcomes that were then refined by the GA, employing selection, crossover, and mutation processes over multiple generations. This approach successfully yielded optimal operating conditions of 34.2099 N load, 0.5006 m/s speed, and 726.7243 m distance, achieving a wear rate of 0.1003×10^{-8} g.N⁻¹.m⁻¹ and a COF of 0.2099, as confirmed through validation tests.

2.3 Additional factors influencing electroless coating performance

2.3.1 Effect of bath loading on coating properties

Genova et al. [36] reported that in pure electroless nickel coatings, an increase in bath loading (defined as the substrate surface area per volume of solution) leads to a higher nickel deposition rate up to a threshold of approximately 1.5 mm²/ml. Beyond this point, the deposition rate tends to plateau, which is attributed to the saturation of the catalytically active sites on the substrate surface. Consequently, further increases in the bath loading did not contribute to the additional growth in the coating thickness.

Similarly, Ergul et al. [117] demonstrated that bath loading has a significant impact on the deposition rate of electroless Ni-P coatings on multiwalled carbon nanotubes (MWCNT). They observed that lower bath loading enhanced the deposition rate, whereas higher bath loading combined with the large surface area of MWCNTs reduced it, emphasizing the need for optimized bath loading to achieve effective metal coatings on MWCNTs.

2.3.2 Antifouling behavior and microbiologically influenced corrosion resistance

Electroless Ni-Cu-P-PTFE coatings improved antifouling performance by incorporating PTFE particles that lowered the surface free energy of the deposits, thereby reducing fouling adhesion on heat exchanger surfaces. The key factor for fouling resistance was the reduced surface free energy rather than the surface roughness, as smoother coatings did not significantly enhance the antifouling properties [118].

The Electroless Ni-B/GO superhydrophobic coating provided antifouling performance by creating a robust micro/nanostructured surface with ultralow water adhesion and a high contact angle (162.8°), enabling water droplets to easily roll off and remove contaminants, as explained by the Cassie-Baxter model. This coating maintained cleanliness even after multiple immersions in muddy water owing to its strong adhesion, mechanical durability, chemical stability in acidic and alkaline solutions, and excellent corrosion resistance, collectively preventing pollutant adhesion and fouling buildup on the AZ91 magnesium alloy surfaces [119].

The electroless NiP-TiNi nanocomposite coating improved the resistance to microbiologically influenced corrosion by exhibiting lower porosity, which limits chloride ion penetration, releases Ni^{2+} ions, and forms TiO_2 , which possesses strong antimicrobial properties against sulfate-reducing bacteria and significantly increases the charge transfer resistance, as shown by electrochemical impedance spectroscopy. These factors together reduce biofilm and corrosion product formation on the surface, leading to an approximately 93% corrosion inhibition efficiency after 28 d of incubation in SRB media [60].

2.4 Research gaps

Based on a literature survey of the different areas of electroless coatings described above, the following research gaps were found that require further investigation.

One of the key areas requiring attention is the development of polyalloy Ni-P coatings incorporating W and Ce. Tungsten has been shown to enhance the hardness, thermal stability, corrosion resistance, and tribological properties of Ni-P coatings. Similarly, the addition of small amounts of Ce to an electroless bath has been shown to improve oxidation resistance, grain refinement, and catalytic properties. However, there is a lack of comprehensive studies exploring the combined effects of tungsten and cerium addition in polyalloy Ni-P coatings, particularly in the context of multifunctional applications.

Current research on Ce-based compounds reveals several gaps, including the limited compatibility of Ce within coating matrices for long-term corrosion protection and an incomplete understanding of its interactions with both the coating and the substrate.

Most existing research has focused on optimizing individual properties in isolation, such as hardness, wear resistance, and corrosion resistance. There is a significant gap in studies that systematically investigate the simultaneous enhancement of the multiple functional properties of electroless coatings. The interplay between various performance factors, such as corrosion resistance, mechanical strength, and tribological behavior, has not been adequately explored.

In addition, advanced optimization techniques, including artificial intelligence (AI) and machine learning (ML), remain underutilized in electroless coating research. These computational tools offer immense potential for improving the process parameters, predicting the coating performance, and identifying optimal compositions more efficiently.

Another critical research gap lies in the elemental analysis of the electroless coating baths. The chemical composition of the coating bath plays a crucial role in determining the final coating properties; however, there is limited research on systematically analyzing and optimizing the bath chemistry to develop high-performance coatings. More studies are needed to understand how variations in bath composition affect the coating quality, stability, and deposition rates.

Moreover, the effects of bath loading, substrate roughness, and activators on electroless coatings have not been reported extensively. These factors significantly impact the coating adhesion, uniformity, and overall performance; however, research addressing their precise effects remains limited. The preparation and surface treatment of substrates before electroless deposition are often overlooked despite being a crucial step in ensuring coating reliability and durability. More systematic studies are needed to investigate the effects of substrate characteristics on the coating microstructure and performance.

2.5 Research objectives

AISI 1010 steel is a low-carbon steel containing approximately 0.1 wt.% carbon, known for its cost-effectiveness and balanced mechanical properties. It has a high Young's modulus of 200 GPa and maximum ultimate tensile strength (UTS) of approximately 650 MPa. Owing to its favorable characteristics, AISI 1010 steel is widely used in applications,

such as automotive components, machinery, and structural elements, where a combination of strength and ductility is essential [120]. The application of electroless coatings to AISI 1010 steel can enhance its corrosion resistance, hardness, wear resistance, and surface uniformity, making it a cost-efficient solution for various industrial applications.

Based on the research gaps stated above, the following objectives were fixed:

- i. Development of multifunctional electroless Ni-W-P-Ce coatings on AISI 1010 steel
- ii. Multi-objective optimization of electroless Ni-W-P-Ce coating parameters
- iii. Investigation of cerium concentration on microstructural evolution and coating performance
- iv. Effect of activator, substrate roughness and bath loading on coating quality

Chapter 3 Development and evaluation of stable electroless Ni-W-P-Ce coating baths

This chapter presents the synthesis and characterization of cerium-added electroless Ni-W-P coatings on AISI 1010 low-carbon steel using four baths with varied compositions and conditions. Mechanical and microstructural analyses were conducted to gain insight into optimizing the coating performance for enhanced durability in demanding applications.

3.1 Experimental details

3.1.1 Substrate preparation

The substrate material used in this study was a 1 mm thick low-carbon steel AISI 1010 sheet procured from the Steel Mart Company, Mumbai, India. The chemical composition of the substrates is listed in Table 3.1. The sheet was sectioned into smaller specimens measuring $15 \times 10 \times 1 \text{ mm}^3$ using wire EDM and subsequently polished with progressively finer grades of silicon carbide sandpaper (400, 800, 1200, 1500, and 2000) to achieve an improved surface finish. To eliminate tightly adhered contaminants, such as dirt, oil, and grease, the substrate surface was initially cleaned with distilled water, followed by acetone in an ultrasonic cleaner (Make: Apex Instruments, Model: UltraS-L20) for 10 min. Next, the substrate was degreased by immersion in a NaOH solution at 40 °C for 15 min. Finally, the activation process was performed by stirring the substrate in 30% diluted HCl solution for 30 s. To prevent contamination by the residual particles, the substrate was thoroughly rinsed with distilled water before and after each step.

Table 3.1. Chemical composition of AISI 1010 steel.

Element	Iron, Fe	Manganese, Mn	Sulphur, S	Phosphorous, P	Carbon, C
Content (%)	99.18-99.62	0.30-0.60	≤ 0.050	≤ 0.040	0.080-0.13

3.1.2 Coating bath preparation

Initially, various bath formulations were developed to deposit Ni-W-P-Ce coatings on low-carbon steel by varying the concentrations of metal salts, complexing agents, and stabilizers, as well as adjusting the bath pH and temperature. After optimization, the four final bath compositions were identified, and the coated samples obtained from these baths were subsequently characterized. The chemical compositions and operating conditions of each bath are listed in Table 3.2.

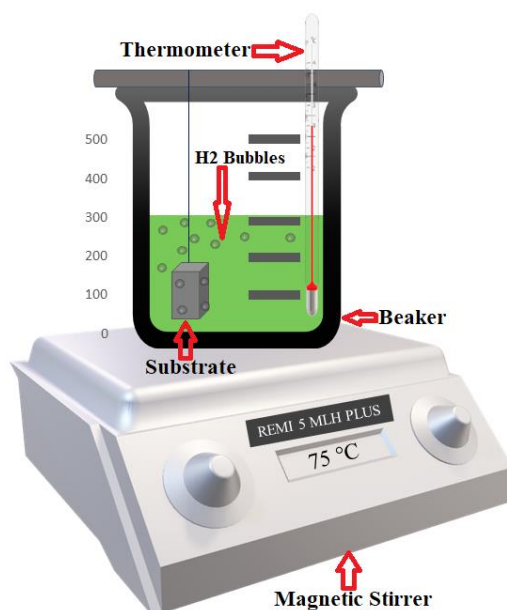


Figure 3.1. Schematic of the electroless coating setup.

3.2 Mechanical and microstructural characterization

After the coating, the rate of deposition was determined by the weight gain method [122] using Equation 3.1.

$$v = \frac{W}{A \times t} \text{ (mg.cm}^{-2} \cdot \text{h}^{-1}\text{)} \quad (3.1)$$

where v is the deposition rate, W is the increment in weight after coating (mg), A is the total surface area of the substrate (cm^2), and t is the coating time (h).

Vickers microhardness measurements were conducted for both the coated samples and substrate using a microhardness tester (UHL VMHT). These measurements ensured that the hardness of the coating layer, rather than the composite hardness, was obtained by determining the indentation depth using the simple relationship given in Equation 3.2. In this equation, ‘ h_i ’ represents the indentation depth, while ‘ d ’ denotes the average diagonal length [123]. During the measurements, the indentation load, indentation speed, turret position, and dwell time were maintained at 50 gf, 30 $\mu\text{m/s}$, 50x magnification, and 10 s, respectively. The indentation load was applied at five different locations on each sample surface, and the average of these five values was considered the final microhardness value for the sample. Additionally, the variation in the diagonal length (ΔD), which represents the tolerance during hardness measurement, was maintained below 5%.

$$h_i = \frac{d}{2\sqrt{2} \tan 68^\circ} = \frac{d}{7} \quad (3.2)$$

The surface morphology and coating composition were observed using field-emission scanning electron microscopy (FESEM) Inspect F50 (FEI, Netherlands). During operation, the high voltage (HV), magnification, working distance (WD), and horizontal field width (HFW) were fixed at 20 kV, 5000x, 12.9 mm, and 59.7 μm respectively. First, highly magnified secondary electron images were captured, and elemental mapping was performed using energy-dispersive X-ray spectroscopy (EDS) coupled with FESEM.

Structural and phase analyses of the coated samples were performed using an X-ray diffractometer (Bruker, D8 Advance) with a scan range of 10-80°. The wavelength of X-rays (λ) in the copper tube was 0.15418 nm.

3.3 Results and discussion

3.3.1 Bath stability

The procedure for preparing the acidic coating bath (Bath A) is illustrated in Figure 3.2. Initially, all metal salts (nickel sulfate and ceric sulfate), except sodium tungstate, were dissolved in distilled water in a beaker, whereas sodium tungstate was separately dissolved in another beaker. All other chemicals, including sodium hypophosphite, sodium acetate, lactic acid, citric acid, and lead(II) acetate, were dissolved in a third beaker, as shown in Figure 3.2(a). Sodium hypophosphite functions as a reducing agent and undergoes oxidation to provide electrons for the reduction process, whereas sodium acetate and lead(II) acetate act as stabilizers, mitigating the rapid pH fluctuations caused by hydrogen ion release during nickel reduction [124]. Acetic acid serves as the primary complexing agent, stabilizing nickel ions and regulating the deposition rate, whereas citric acid, acting as an auxiliary complexant, enhances the bath stability but reduces the coating speed at higher concentrations [125]. Separate dissolution of these components is crucial for preventing premature reactions between the metal ions and reducing agents. In electroless coatings, the reduction of metal ions by a reducing agent is the primary reaction requiring strict control. If metal salts and reducing agents are directly mixed as dry components or in concentrated solutions, immediate reactions can occur, leading to uncontrolled metal reduction outside the coating process [126]. Once all chemicals were fully dissolved, they were combined in a single beaker, as depicted in Figure 3.2(b).

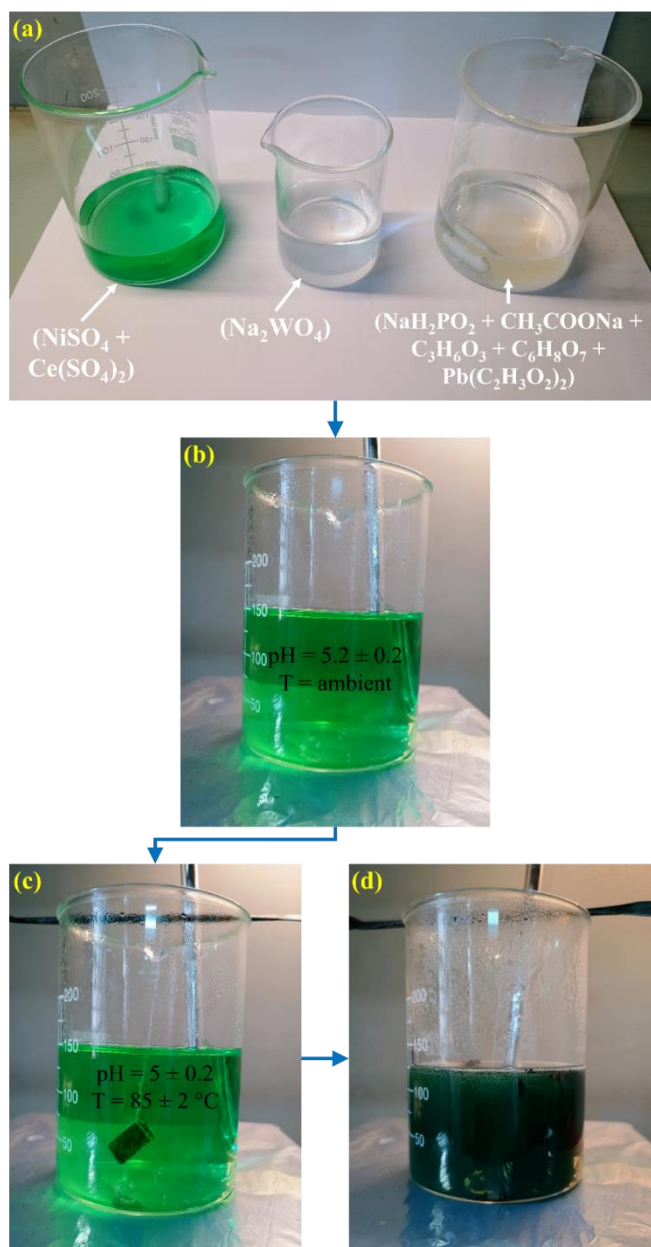


Figure 3.2. Sequential steps for preparation of Bath A.

At this stage, the bath appeared clear and transparent, with a light green color and pH of approximately 5.2. The bath temperature was increased to $85 \text{ }^\circ\text{C}$ before immersing the substrate, as shown in Figure 3.2(c). During this step, the bath remained transparent and the pH stabilized at approximately 5.

The coating process was performed for approximately one hour. Subsequently, spontaneous bath decomposition was observed, marked by rapid hydrogen bubble generation and a color change from green to black, as illustrated in Figure 3.2(d). The acidic nature of this bath and elevated coating temperature ($85 \text{ }^\circ\text{C}$) contributed to its

decomposition. This instability can be attributed to the chemical degradation of reducing agents, such as hypophosphite, under acidic conditions, where they tend to oxidize or decompose prematurely. Acidic baths promote a higher rate of hydrogen evolution, which adversely affects the coating reaction by locally depleting active species. Additionally, elevated temperatures accelerate side reactions, further destabilizing the bath [127].

Another basic bath (Bath B) was prepared, which decomposed within just half an hour. This was likely due to the absence of a stabilizer and high concentration of cerium. A high Ce ion concentration promotes particle agglomeration, which blocks catalytically active sites on the substrate surface and increases the surface energy of the coating, thereby inhibiting nucleation and growth. Moreover, the excessive presence of Ce ions destabilizes the coating bath, resulting in a reduced deposition efficiency [121].

A neutral electroless coating bath (Bath N1) was formulated using fewer chemicals than Bath A, which contained two complexing agents. Bath N1 utilizes only one trisodium citrate. Sodium citrate, known for its moderate complexing ability, helps stabilize the coating solution by forming stable complexes with Ni^{2+} ions [128]. This stabilization was essential for maintaining the desired pH and chemical environment during the coating process. Additionally, instead of using two stabilizers, as in Bath A only one stabilizer, thiourea, was used in this formulation. In acidic electroless nickel baths, thiourea plays a critical role in preventing the formation of nickel phosphite precipitates when sodium hypophosphite is used as a reducing agent. Similarly, in alkaline baths, thiourea ensures the solubility of nickel ions, preventing their precipitation as nickel hydroxide and thereby enhancing the overall stability of the coating solution [129].

To prepare Bath N1, all metal salts were dissolved in a beaker with distilled water, whereas the remaining chemicals were dissolved separately in another beaker, as illustrated in Figures 3.3(a) and 3.3(b). After the chemicals were dissolved separately, they were combined in a single beaker, as shown in Figure 3.3(c). Because all the metal salts were dissolved together in a single beaker, the solution appeared nontransparent. At this stage, the solution remained nontransparent with a light blue appearance, and the pH was measured at approximately 7.2. Upon heating the bath to 75 °C, the color of the solution changed from nontransparent light blue to slightly transparent green, as shown in Figure 3.3(d). During this process, pH decreased to 6.2. This decrease in pH with increasing temperature is primarily attributed to the enhanced ionization of water, which increases the

concentration of hydrogen ions (H^+), thereby lowering the pH. Additionally, elevated temperatures influence the solubility of gases and chemical species in the bath, promoting reactions that generate acidic by-products or deplete hydroxide ions (OH^-), further reducing the solution pH [130]. It is important to note that the coating temperature was maintained at $75\text{ }^\circ\text{C}$ to avoid issues encountered at higher temperatures in Bath A.

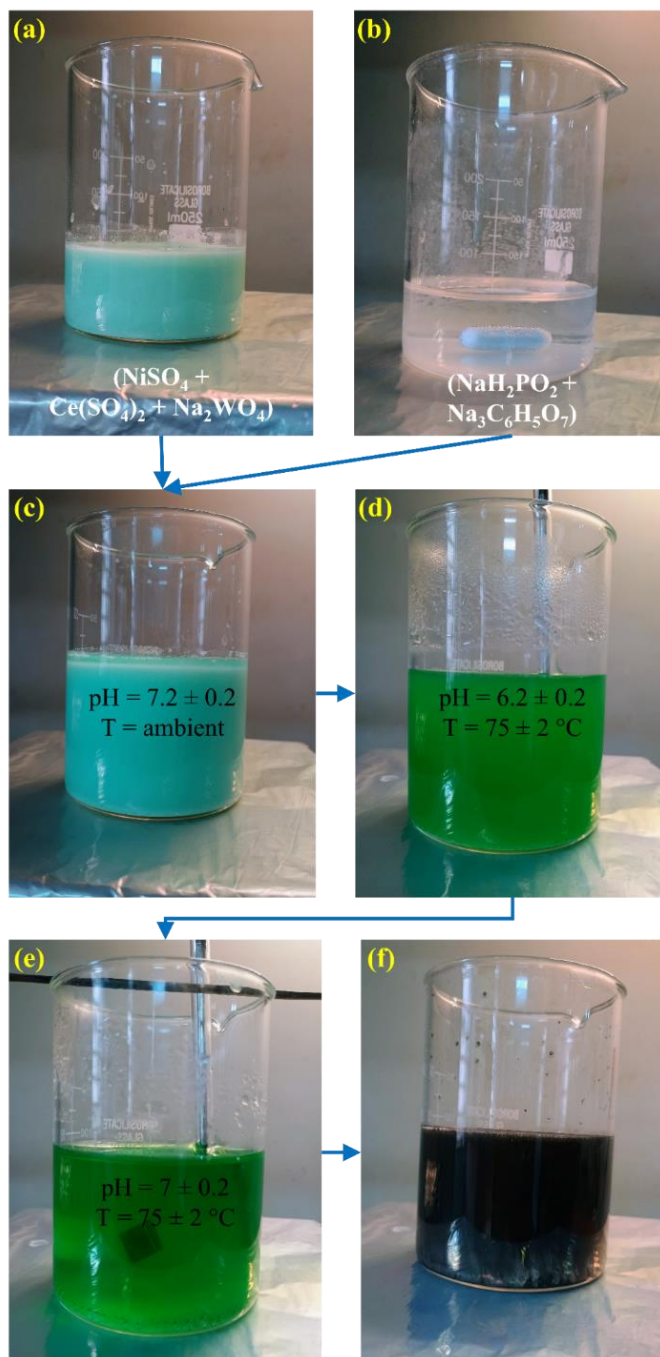


Figure 3.3. Preparation and evolution of Bath N1.

Subsequently, a small amount of ammonia solution was added to the bath, which increased the pH to approximately 7 and further improved the transparency of the solution, as shown in Figure 3.3(e). The addition of ammonia to electroless coating baths serves multiple functions, including the formation of stable metal-ammonia complexes that enhance the solubility of metal ions, facilitating their reduction, and regulating the bath pH, which is crucial for optimizing the deposition conditions. Furthermore, ammonia stabilizes the coating bath, prevents precipitation, and ensures uniform deposition [131]. However, even after this adjustment, the transparency of the bath was not as high as that of Bath A.

The coating process was then performed for one hour, during which the deposition rate was significantly higher than that in Bath A. This aspect is discussed in Section 4.4.2. However, the bath eventually decomposes, as shown in Figure 3.3(f). It appears that tungsten dissolution issues persisted, likely because of its direct mixing with other metal salts under ambient conditions. A white precipitate, most likely tungsten, was observed at the bottom of the beaker, contributing to the decomposition of the bath.

A second neutral bath (Bath N2) was prepared to address the bath decomposition observed in Bath N1. The preparation procedure and chemical composition were identical to those used for Bath N1, as shown in Figure 3.4. The only modification was the dissolution of the sodium tungstate. Sodium tungstate was first dissolved in distilled water in a separate beaker instead of mixing it directly with other metal salts under ambient conditions. It was then introduced into a bath containing all other chemicals once the bath temperature reached 55 °C, as illustrated in Figure 3.4(d). At this stage, pH was maintained at 6.5. Subsequently, ammonia solution was added to the bath, the pH was increased to 7.5, and the final bath temperature was raised to 75 °C. With this modified preparation method, all the chemicals were completely dissolved, resulting in a highly transparent solution, as shown in Figure 3.4(e). The deposition rate was satisfactory. The bath remained stable for up to two hours; however, after this period, it became significantly more viscous, as shown in Figure 3.4(f). The increase in viscosity over time in the electroless coating baths is primarily attributed to the accumulation of salts and byproducts such as sulfates and phosphites, which form during the coating process. This build-up disturbs the chemical equilibrium of the solution, leading to an increased viscosity. These factors ultimately affect the quality and performance of the coatings [132].

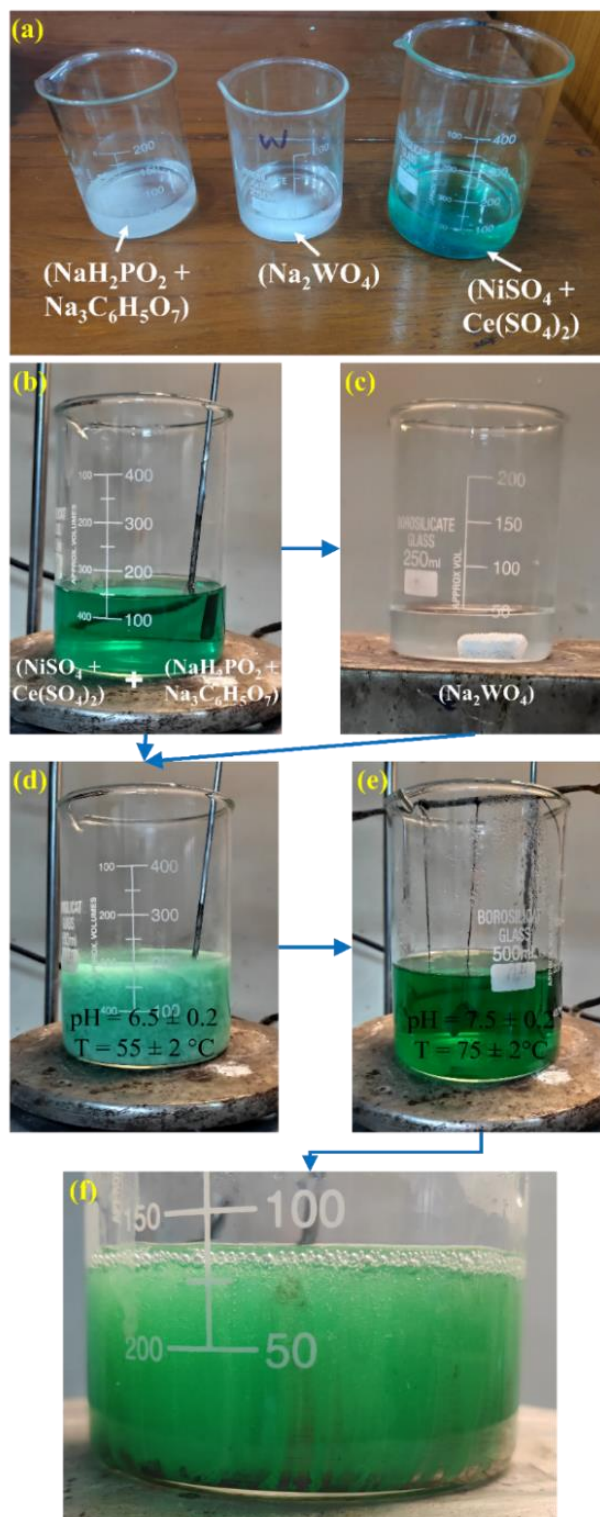


Figure 3.4. Preparation and stability assessment of Bath N2.

3.3.2 Deposition rate

The deposition rates for all the four baths are shown in Figure 3.5. The 1st Bath A, which contained numerous chemicals, exhibited the lowest deposition rate. When the pH is low

(high concentration of H^+ ions), reactions producing nickel from nickel ions (Ni^{2+}) and the reducing agent ($H_2PO_2^-$) are less favorable. Excess H^+ ions drive the reaction backward, resulting in fewer Ni atoms being deposited onto the substrate. Additionally, the slow deposition rate in the acidic medium may be due to the reduced bath stability and formation of insoluble species such as $NiHPO_3$, which precipitates out of the solution and consumes nickel ions, further hindering the deposition process [133]. This bath also had a relatively high phosphorus content, as shown in Table 3.3. A higher phosphorus content typically correlates with a slower deposition rate, regardless of the bath type (acidic or alkaline). This is due to several interrelated factors: the formation of complex species that do not readily participate in the deposition reaction, alteration of the reduction potential of nickel ions, and impeded adsorption of nickel ions on a surface saturated with phosphorus [109,134].

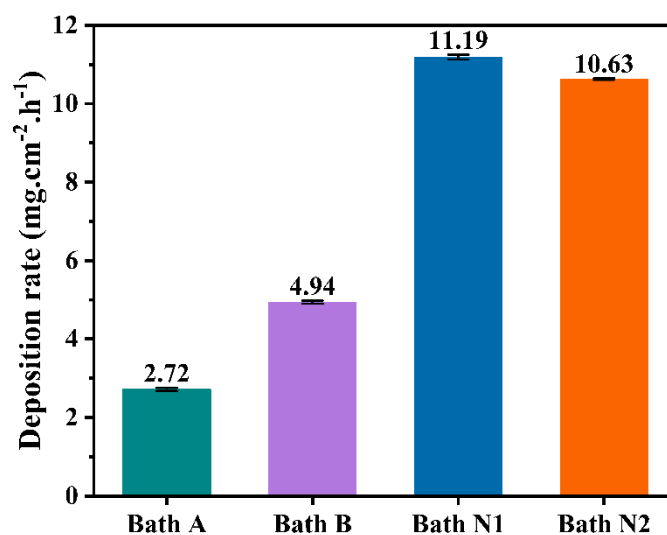


Figure 3.5. Deposition rate of the Ni-W-P-Ce coatings.

Bath B also showed a low deposition rate, primarily because of the absence of a stabilizer and high concentration of cerium. A high Ce ion concentration leads to particle agglomeration, blocking the catalytic active sites on the substrate surface and increasing the surface energy of the coating, which hinders nucleation and growth. Additionally, the presence of excess Ce ions destabilized the coating bath, leading to a decreased deposition efficiency [121].

Bath N1 and N2, which included the stabilizer thiourea, exhibited higher deposition rates, with Bath N1 exhibiting rate of $11.19 \text{ mg.cm}^{-2} \cdot \text{h}^{-1}$. Thiourea prevents the complexing agent sodium citrate by effectively controlling the release rate of free Ni ions during the reduction reaction, which can increase the deposition rate [135]. In these baths, an optimal

amount of cerium salt is effectively adsorbed onto the substrate surface, reducing the surface energy and facilitating nickel ion reduction, thereby accelerating the coating process [136]. Bath N1, which had a slightly lower pH, decomposed after 1 h of coating, whereas Bath N2, which had a pH of approximately 7.5, remained stable and showed no issues.

3.3.3 Microhardness

The average Vickers microhardness measured at 50 gf for both the substrate and the coated samples is shown in Figure 3.6. The microhardness values of the samples obtained from Bath A and Bath B were more than twice that of the substrate. In contrast, the samples from Bath N1 and Bath N2 demonstrated microhardness values more than four times those of the substrate. The lower microhardness observed in Bath A and Bath B was attributed to the lower deposition rates, as discussed in the deposition rate subsection of the Results and Discussion section. This is corroborated by the SEM images in Figure 3.7, where Bath A shows nonuniform deposition with coating nodules visible only in certain areas of the microstructure. Similarly, the SEM images of Bath B revealed incomplete coverage of the substrate surface by the coating alloys, leading to significant variance and lower microhardness values.

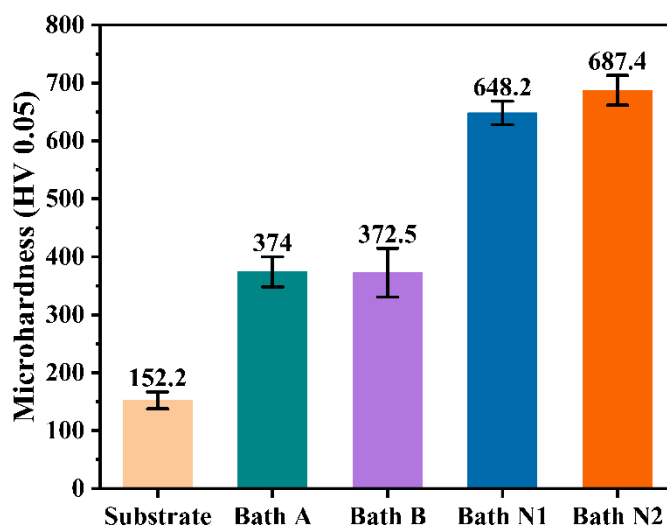


Figure 3.6. Vickers microhardness of the substrate and Ni-W-P-Ce coatings.

However, the coatings produced from Bath N1 and Bath N2 exhibit more compact microstructures. The substantial increase in the microhardness of these samples was primarily due to the presence of small amounts of phosphorus, as indicated by the EDS

results in Table 3.3. A low phosphorus content induces higher internal stresses in the coating, thereby increasing microhardness. Additionally, the formation of a solid solution of tungsten within the nickel matrix contributed to an increase in microhardness. The use of Ce in these baths further refines the microstructure. The hardening effects of Ce as a second phase, enhanced nucleation rates, and stability of the coating under alkaline conditions also play a significant role in increasing microhardness [137,138].

3.3.4 Surface morphology and elemental composition

The surface morphologies of the samples coated with the four baths are shown in Figure 3.7, with each exhibiting distinct characteristics. The surface obtained from Bath A (Figure 3.7(a)) shows a very poor morphology owing to minimal deposition. Coating nodules are sparsely distributed and only visible in a few areas, as indicated by the yellow arrows.

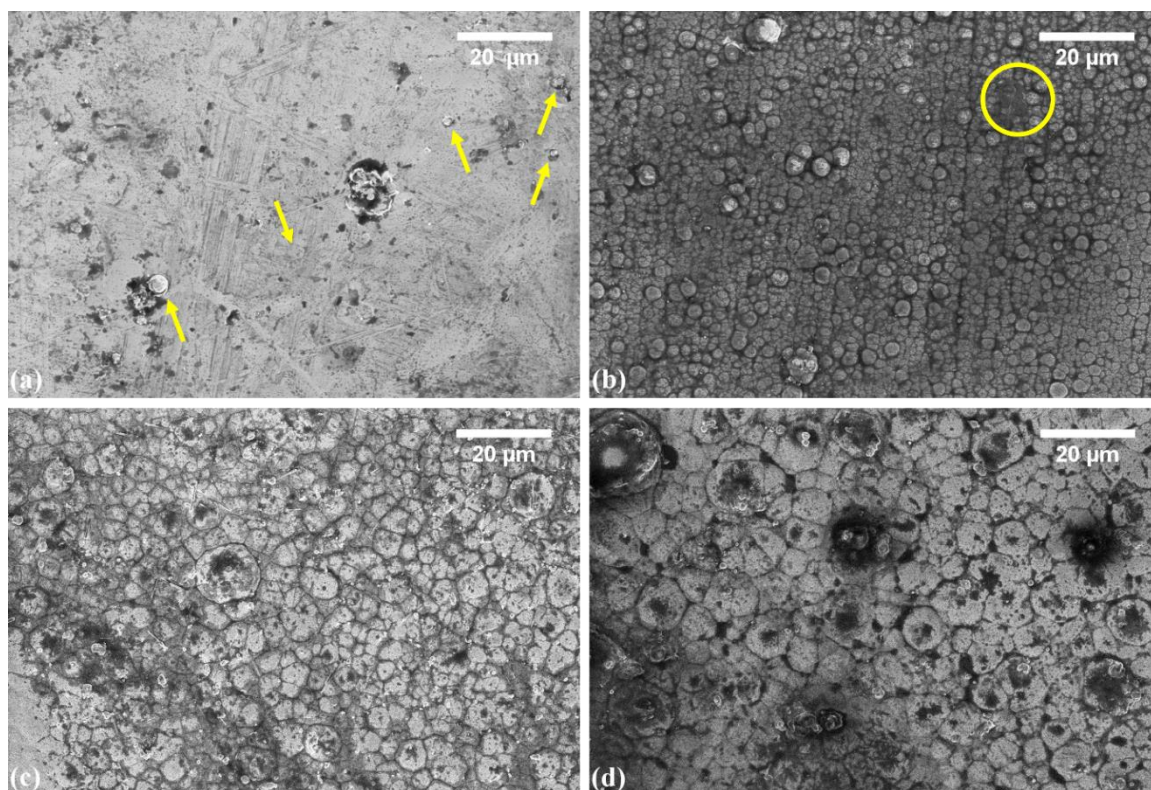


Figure 3.7. Surface morphology of Ni-W-P-Ce coatings: (a) Bath A; (b) Bath B; (c) Bath N1; (d) Bath N2.

In contrast, the surface (Figure 3.7(b)) obtained from Bath B displays visible globular or spherical nodules, which are typical of electroless coatings. This nodule formation results from a rapid nucleation process during electroless deposition, where small clusters

of atoms form quickly, limiting surface diffusion and causing atoms to accumulate into nodules rather than spread evenly. This type of morphology is also influenced by factors such as the bath composition and the condition of the substrate surface [139]. The surfaces obtained from Bath N1 and Bath N2 (Figure 3.7(c) and (d)) exhibited similar spherical or cauliflower-like microstructures. The average nodule sizes for the samples from the Bath B, Bath N1, and Bath N2 were 0.91 μm , 2.05 μm , and 2.07 μm , respectively. The smaller nodule size in Bath B can be attributed to two factors. First, the coating time was brief, preventing the full growth of the nodules, as some regions remained nodule-free, as highlighted by the yellow circle. Second, the high concentration of cerium sulfate in this bath played a significant role. Cerium particles acted as nucleation sites during the coating process, promoting the formation of new grains rather than the enlargement of existing grains, leading to a finer microstructure with smaller nodules [106]. Notably, none of the four coated samples exhibited any cracks, indicating good structural integrity across all coatings.

The weight percentages of the elements present in all the four samples are listed in Table 3.3. Bath A sample, with 8.5 wt.% phosphorus (P), represents a medium phosphorus (6-9 wt.%) electroless coating, which is known for its high corrosion and abrasion resistance. Bath B sample, containing 10.2 wt.% P, falls into the category of high phosphorus (10-15 wt.%) electroless coatings, which are typically associated with good ductility and enhanced corrosion resistance, particularly against chlorides.

Table 3.3. Elemental compositions (wt.%) of coatings deposited from different coatings bath.

Element	Bath A	Bath B	Bath N1	Bath N2
Nickel, Ni (wt.%)	76.2	73.1	78.9	78.8
Tungsten, W (wt.%)	14.5	15.9	16.0	17.6
Phosphorous, P (wt.%)	8.5	10.2	4.0	2.8
Cerium, Ce (wt.%)	0.8	0.9	1.1	0.8

In contrast, the Bath N1 and Bath N2 samples had lower phosphorus contents of 4 wt.% and 2.8 wt.%, respectively. These values represent low phosphorus (3-5 wt.%) electroless coatings, which are generally characterized by superior wear resistance [35,140]. Figures 3.8, 3.9, 3.10, and 3.11 show the elemental mapping for all elements in the four samples, illustrating the uniform distribution of elements across the coated surfaces. The images

clearly demonstrate the dense distribution of phosphorus in the medium- and high-phosphorus coatings, whereas a noticeable decrease in the phosphorus content was observed in the low-phosphorus coatings. Additionally, Ce was uniformly present in all the samples, with the highest content of 1.1 wt.% observed in Bath N1 sample. This confirms the successful integration of Ce into the electroless Ni-W-P coatings. Moreover, the tungsten content was consistently high across all four samples, exceeding 10 wt.%. The presence of tungsten contributes to solid-solution strengthening within the nickel matrix and forms a passivation film of WO_3 , which enhances both the hardness and corrosion resistance of the coatings [128].

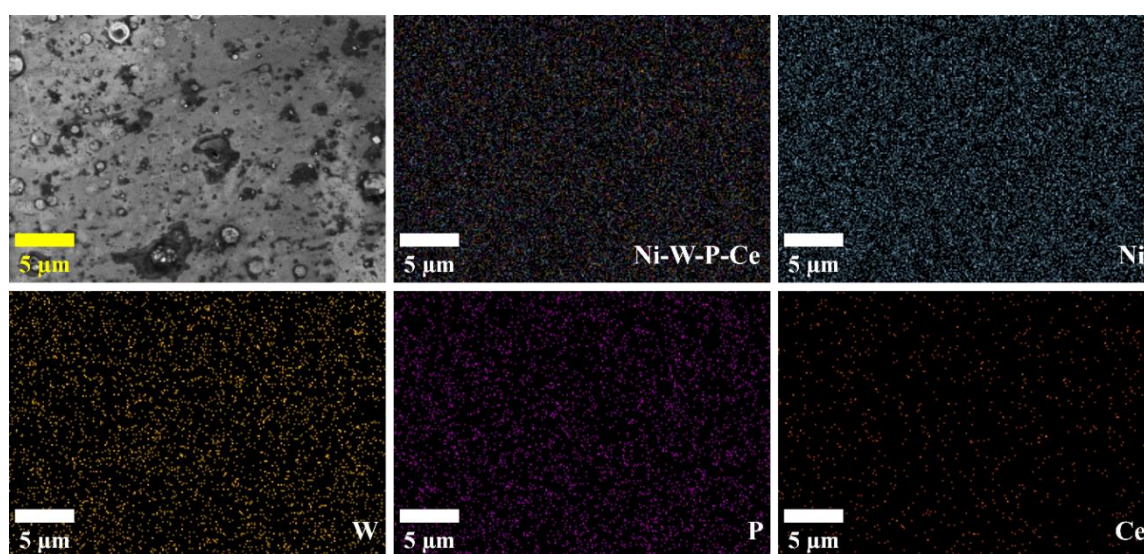


Figure 3.8. EDX mapping images of the coating top surfaces in Bath A.

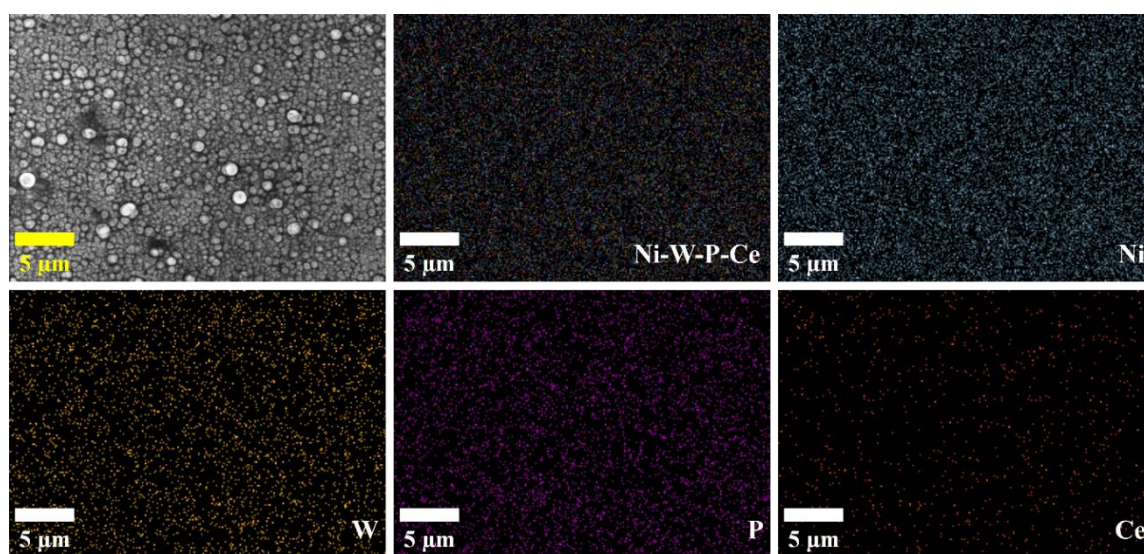


Figure 3.9. EDX mapping images of the top surfaces of the coatings in Bath B.

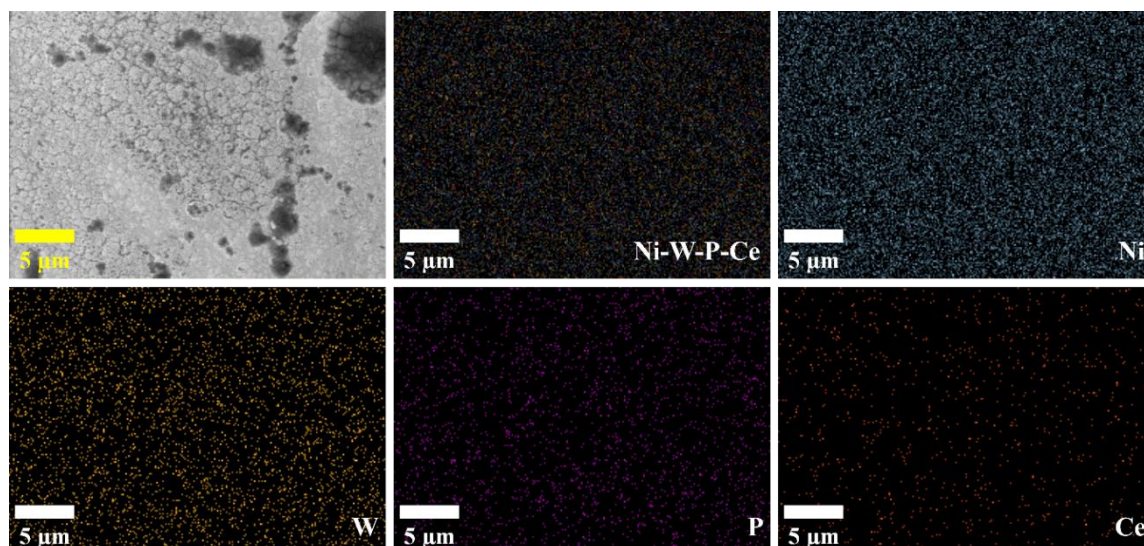


Figure 3.10. EDX mapping images of the coating top surfaces of Bath N1.

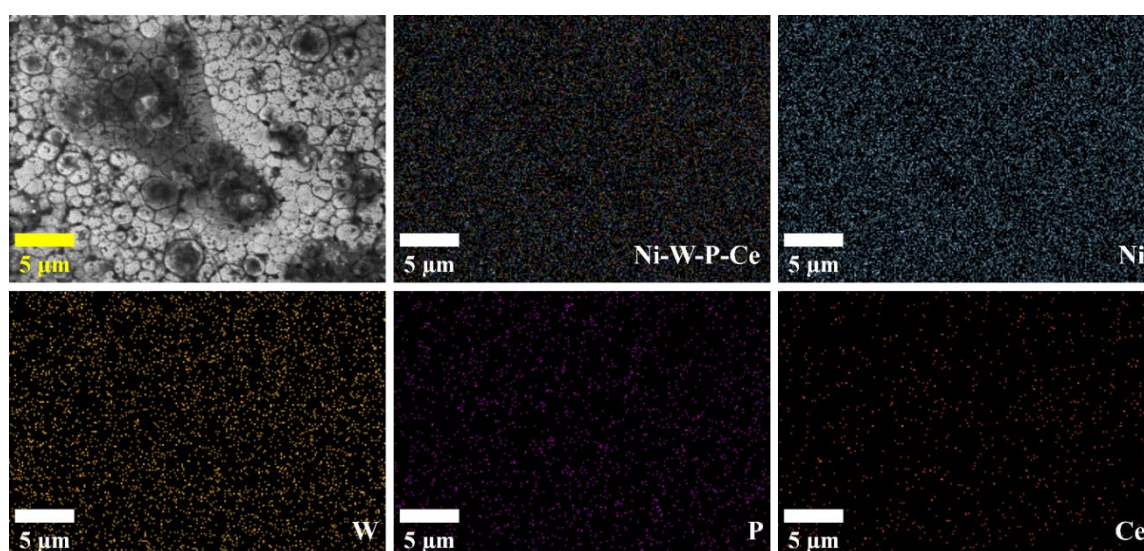


Figure 3.11. EDX mapping images of the coating top surfaces in Bath N2.

3.3.5 Coating structure

Figure 3.12 shows the X-ray diffraction (XRD) patterns of all four samples, revealing similar profiles with variations in peak intensity and broadness. The Ni peak at approximately 44.2° shows a lower intensity and narrower width for the Bath A and Bath B samples than for the Bath N1 and Bath N2 samples, which exhibit a higher peak intensity and broader peaks. The peak intensity is influenced by the atomic positioning within the unit cell, whereas the peak broadness is related to the crystallite size and lattice strain. Crystallites are uniform regions within polycrystalline solids that maintain consistent crystal structures [141]. The higher peak intensity and broader peaks in the Bath N1 and Bath N2 samples suggest a greater degree of crystallinity or a larger amount of the nickel

crystal phase compared to the Bath A and Bath B samples. Additionally, the broader peaks in the Bath N1 and Bath N2 samples indicated smaller crystallite sizes.

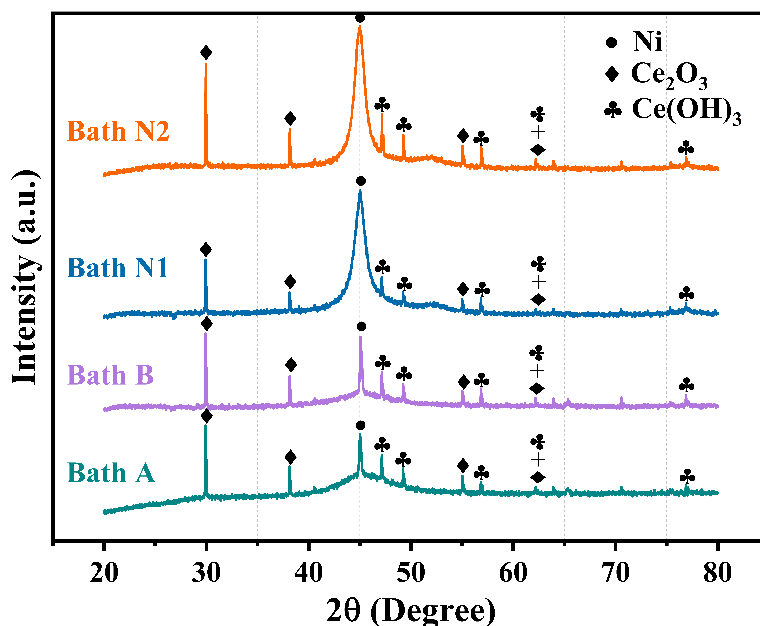


Figure 3.12. XRD patterns of all 4 samples.

In addition to the Ni peaks, the XRD patterns also show peaks corresponding to Ce_2O_3 and $\text{Ce}(\text{OH})_3$. These peaks were identified using Match software with entry numbers 96-210-0638 for nickel, 96-153-1456 for cerium oxide, and 96-210-7093 for cerium hydroxide, in agreement with previous studies [54,105,142]. The formation of cerium hydroxide and subsequent cerium oxide during the drying process was observed, with cerium ions initially reacting with hydroxide ions to form cerium hydroxide, which was later converted to cerium oxide [143].

3.4 Closure

This chapter describes the development and evaluation of stable electroless Ni-W-P-Ce coating baths for depositing coatings on low-carbon steel AISI 1010. Four coating baths with different chemical compositions and operating conditions are developed. Bath A and Bath B exhibited low deposition rates and poor coating quality owing to their instability and high Ce content. Bath N1 and Bath N2, which contained thiourea, showed higher deposition rates and better coating quality. Bath N2 with a pH of 7.5 and bath temperature of 75 °C yielded the best results. Microhardness testing revealed that the Bath N1 and Bath N2 samples had hardness values more than four times higher than that of the substrate,

which was attributed to the presence of phosphorus, tungsten, and cerium. The SEM analysis showed distinct surface morphologies for each bath, with the Bath N1 and Bath N2 samples exhibiting compact nodular microstructures. EDS confirmed the uniform distribution of elements including Ce across the coated surfaces. The XRD analysis revealed the presence of nickel, cerium oxide, and cerium hydroxide phases, with the Bath N1 and Bath N2 samples displaying higher crystallinity. Based on mechanical and microstructural characterization, the chemical composition of Bath N2 was found to be the most advantageous for achieving superior coating performance.

Chapter 4 Multi-objective optimization of electroless Ni-W-P-Ce coating bath compositions

This chapter provides a detailed comparison of various optimization techniques for electroless Ni-W-P-Ce coating bath composition, highlighting the integration of artificial neural networks with metaheuristic algorithms to improve prediction accuracy and optimization efficiency.

4.1 Experimental procedure and parameter selection

The chemicals and operating conditions used for the coatings are presented in Table 4.1. The concentration range of the four primary coating element salts was selected based on the data from Chapter 3, while all other operating conditions remained unchanged.

Table 4.1. Chemical composition and operating conditions.

Chemical	Concentration	Conditions
NiSO ₄ .6H ₂ O	15-40 g/L	Temp.: 75 ± 2 °C
Na ₂ WO ₄ .2H ₂ O	15-30 g/L	pH: 7.5 ± 0.2
Ce(SO ₄) ₂ .4H ₂ O	4-12 mg/L	Time: 2 hours
NaH ₂ PO ₂ .H ₂ O	25-40 g/L	
Na ₃ C ₆ H ₅ O ₇ .2H ₂ O	20 g/L	
CH ₄ N ₂ S	0.001 g/L	

Electroless Ni-W-P-Ce coating was applied on plain carbon steel following the design created through central composite design of response surface methodology. In this design, the input variables or parameters were the amounts of the four main coating elements: Ni, W, P, and Ce. The pilot experiments showed that the concentration ranges of the chemicals listed in Table 4.2 had the most significant influence on the effectiveness of the coatings. The ±2 coded levels represent the axial (star) points.

After coating, the effects of these variables were measured for five responses: surface roughness, microhardness, fouling weight, surface free energy, and corrosion protection efficiency. A circumscribed central composite design of the response surface methodology with these parameters was constructed using MINITAB 22 software, considering the default distance α of the axial points from the center to be 2. Thirty experimental runs were conducted.

Table 4.2 Input parameters and their levels.

Factors	Units	Code	Levels				
			-2	-1	0	1	2
Nickel sulfate ($\text{NiSO}_4 \cdot 6\text{H}_2\text{O}$)	g/L	Ni	2.5	15	27.5	40	52.5
Sodium hypophosphite ($\text{NaH}_2\text{PO}_2 \cdot \text{H}_2\text{O}$)	g/L	P	17.5	25	32.5	40	47.5
Sodium tungstate ($\text{Na}_2\text{WO}_4 \cdot 2\text{H}_2\text{O}$)	g/L	W	7.5	15	22.5	30	37.5
Cerium sulfate ($\text{Ce}(\text{SO}_4)_2 \cdot 4\text{H}_2\text{O}$)	mg/L	Ce	0	4	8	12	16

4.2 Response measurement

The surface roughness of the coated samples was measured by using a Mitutoyo SJ-210 contact-type surface profilometer. The arithmetical mean roughness value (R_a) was recorded with a cutoff length (λ_c) of 0.8 mm and a measuring speed of 0.5 mm/s, as per ISO 1997 standards. Five measurements were taken at different locations on each sample and the average was calculated as the final R_a value.

The microhardness values of the coatings were measured using a Vickers microhardness tester (UHL VMHT). A load of 50 gf was applied with a dwell time of 12 s and speed of 25 $\mu\text{m/s}$. Similar to the surface roughness measurements, five readings were obtained for each specimen and the average was used as the final microhardness value.

To evaluate the antifouling performance, the coated samples and substrate were boiled in normal tap water with a total dissolved solids (TDS) value of 2460 ppm for 20 h at the Surface Engineering Laboratory at Jadavpur University, Kolkata, following procedures adopted by previous researchers [118,144]. Samples were collected at 4-hour intervals, dried, and weighed using a precision balance with a resolution of 0.1 mg. The fouling rate was determined based on the weight gain due to the fouling deposits. The surface morphology of the adhered fouling was examined using FESEM in the backscattered imaging mode, and elemental mapping of the deposits was performed using EDS.

The surface free energy of the solid samples was calculated using the Owens and Wendt geometric mean approach, which considers the dispersion and polar components of surface energy. The dispersion energy is associated with Van der Waals forces, whereas the polar energy corresponds to interactions, such as hydrogen bonding. Contact angles of two test liquids, distilled water ($\gamma_L=72.8$ mN/m, $\gamma^d=21.8$ mN/m, $\gamma^p=51$ mN/m) and ethylene glycol ($\gamma_L=48$ mN/m, $\gamma^d=29$ mN/m, $\gamma^p=19$ mN/m), were measured on the coated and substrate

samples using a KYOWA Interface Measurement and Analysis System (FAMAS). The average of three contact angle measurements at room temperature (300 K) was used to calculate the surface free energy using Equations 4.1 and 4.2 [145].

$$y_L(1 + \cos \theta) = 2(\sqrt{y_S^d y_L^d} + \sqrt{y_S^p y_L^p}) \quad (4.1)$$

$$y_S = y_S^d + y_S^p \quad (4.2)$$

Where y_L and y_S are the surface tension of the test liquid and surface free energy of the solid surface, respectively. y^d and y^p are the dispersion and polar components, respectively. θ is the contact angle between the liquid and solid surfaces.

Electrochemical measurements were conducted using a Metrohm μ Stat i400 potentiostat in a standard three-electrode setup comprising an Ag/AgCl reference electrode, platinum counter electrode, and electroless Ni-W-P-Ce coated sample as the working electrode. A circular area with a diameter of 10 mm was exposed to a 3.5% NaCl solution as a corrosive medium. Open circuit potential (OCP) measurements were performed for 1 h followed by linear potentiodynamic polarization (LP) tests. LP tests were conducted by polarizing the working electrode in the cathodic and anodic directions within ± 250 mV of the OCP, at a scanning rate of 2.5 mV/s. The Dropview software (Metrohm, India) was used for data acquisition and analysis. The protection efficiency (PE) of the deposited coatings was calculated from the Tafel plots by finding the ratio of corrosion current density of the substrate (I_{cor}^S) to that of the coated surface (I_{cor}^C) using the relationship (Equation 4.3) [146]-

$$PE = 1 - \frac{I_{cor}^C}{I_{cor}^S} \quad (4.3)$$

4.3 Optimization using RSM

Response surface methodology (RSM) employs desirability functions [147] to optimize multiple responses by transforming each response into desirability scores ranging from 0 to 1, where 1 indicates an ideal outcome and 0 signifies an unacceptable result (Equation 4.4). These functions are tailored based on objective maximization, minimization, or targeting a specific value. For maximization, the desirability (d_i) increases linearly from a

lower threshold (L) to the target (T) and remains at 1 beyond T, whereas for minimization, the function is inverted.

$$d_i = \begin{cases} 0 & \text{if } y_i < L \\ \frac{y_i - L}{T - L} & \text{if } L \leq y_i \leq T \\ 1 & \text{if } y_i > T \end{cases} \quad (4.4)$$

Where y_i is the response variable of interest for the i -th response. The overall desirability (D) for multiple responses is calculated using the geometric mean of the individual desirability scores (Equation 4.5), ensuring that the overall desirability is zero if any response is undesirable.

$$D = (d_1 \cdot d_2 \dots d_m)^{1/m} \quad (4.5)$$

Where ‘m’ is the number of responses. Optimization involves maximizing D by adjusting the design variables using numerical methods such as gradient-based algorithms or direct search techniques. This approach effectively balances conflicting objectives by assigning weights to prioritize certain responses, enabling a trade-off between competing goals such as maximizing yield while minimizing cost. Visualization tools, such as contour and surface plots, aid in interpreting results, showing how desirability changes across the design space, and highlighting optimal performance regions. This methodology facilitates a systematic and flexible approach for multi-objective optimization.

4.4 Optimization using hybrid ANN-based metaheuristic algorithms

Multi-objective optimization was performed using three hybrid ANN-based metaheuristic algorithms: teaching learning-based optimization (TLBO), whale optimization algorithm (WOA), and red fox optimization (RFO), as illustrated in the flowchart (Figure 4.1). A weighted sum method was used to convert multiple objectives into a single scalar objective, using predetermined weights. This method simplifies the optimization process by combining various objectives into a single-value function, allowing the use of standard optimization techniques. In surface engineering, scalarization aids material selection by balancing properties, such as thickness, adhesion, and cost [148].

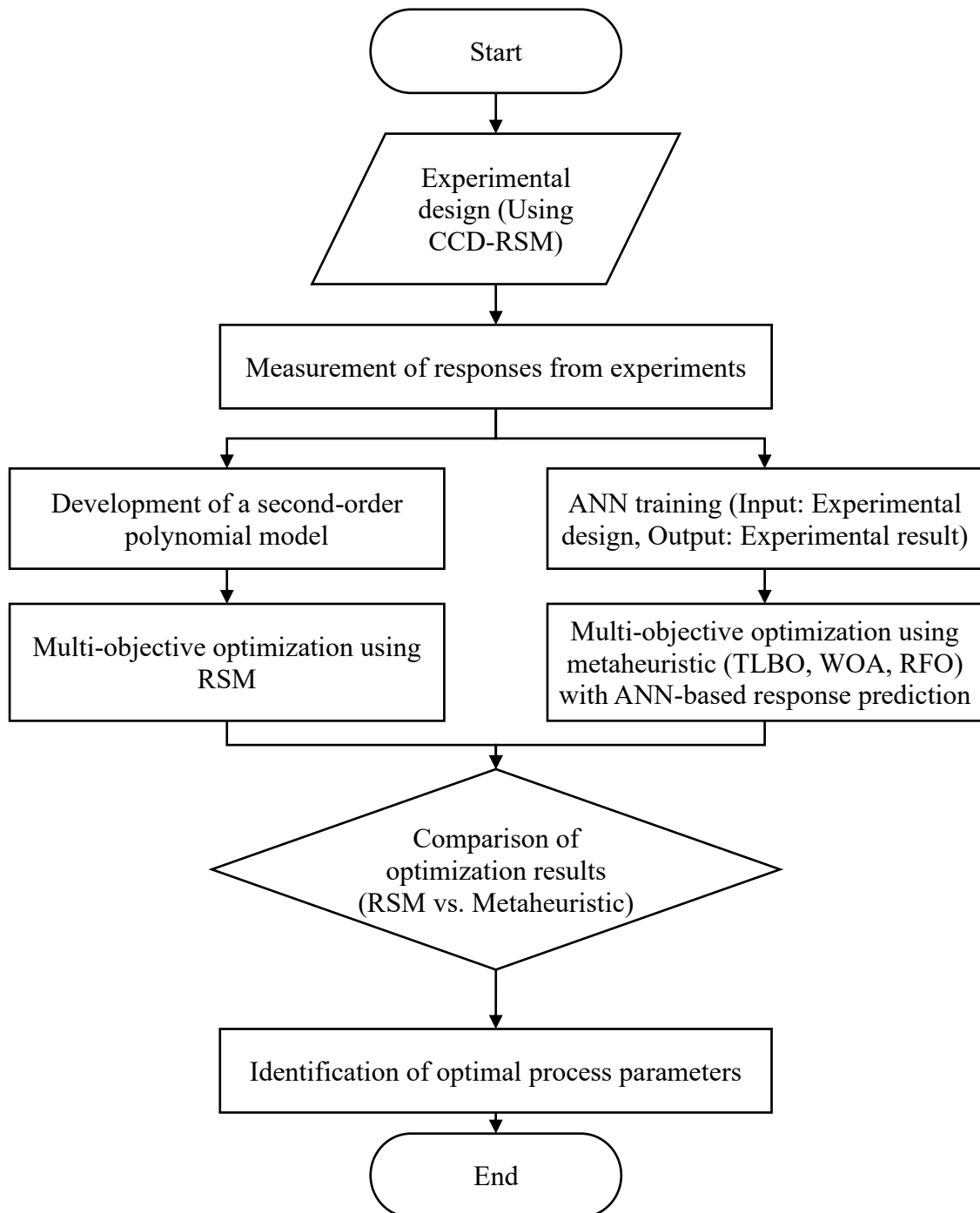


Figure 4.1. Flowchart of the optimization process for the Ni-W-P-Ce coatings.

The problem formulation is as follows:

Weights: $w = (w_1, w_2, \dots, w_m)$,

$$\sum_{i=1}^m w_i = 1 \quad (4.6)$$

$$\text{Minimize (or Maximize)} F_{\text{weighted}}(X) = \sum_{i=1}^m w_i f_i(X) \quad (4.7)$$

where w_i is the weight assigned to the i -th objective function. The weights indicate the relative importance of each objective, which is typically based on the preferences of the decision-maker. These weights dictate the trade-offs between objectives, and varying them can lead to different optimal solutions, thereby facilitating exploration of the Pareto front [149,150].

The ANN trained on the RSM-designed experimental data was integrated into all the three algorithms. Unlike conventional approaches that rely on second-order objective functions to calculate the response values within optimization algorithms, this hybrid implementation utilizes the ANN model to compute the response values. This innovation enhances parameter optimization and response prediction in complex engineering problems where traditional mathematical modeling poses challenges. The hybrid methodologies ANN-TLBO, ANN-WOA, and ANN-RFO were implemented programmatically in MATLAB R2023a. The complete MATLAB script used for conducting the multi-objective optimization is provided in the Appendix. Population sizes of 100 and 50 iterations were chosen to evaluate the consistency of the results across all the three hybrid algorithms.

The ANN employed in this study was a cascade-forward neural network [151], characterized by its feedforward connections and additional direct links between the input and subsequent layers. The architecture of the cascade-forward ANN is shown in Figure 4.2. The network architecture comprises five hidden layers with sizes of 15, 12, 10, 8, and 6 neurons, respectively. The input and output layer sizes were not explicitly specified in the training script because they were determined using the dataset used. The training algorithm utilized is the Levenberg-Marquardt ('trainlm') algorithm, which is well-suited for function approximation and regression tasks due to its efficiency and robustness. The activation functions used in the network include the hyperbolic tangent (tansig) function for the hidden layers, which introduces nonlinearity and enables the network to model complex relationships, and the linear (purelin) function for the output layer, which ensures compatibility with regression outputs. Input and output data are preprocessed using the 'removeconstantrows' and 'mapminmax' functions to enhance training stability and performance.

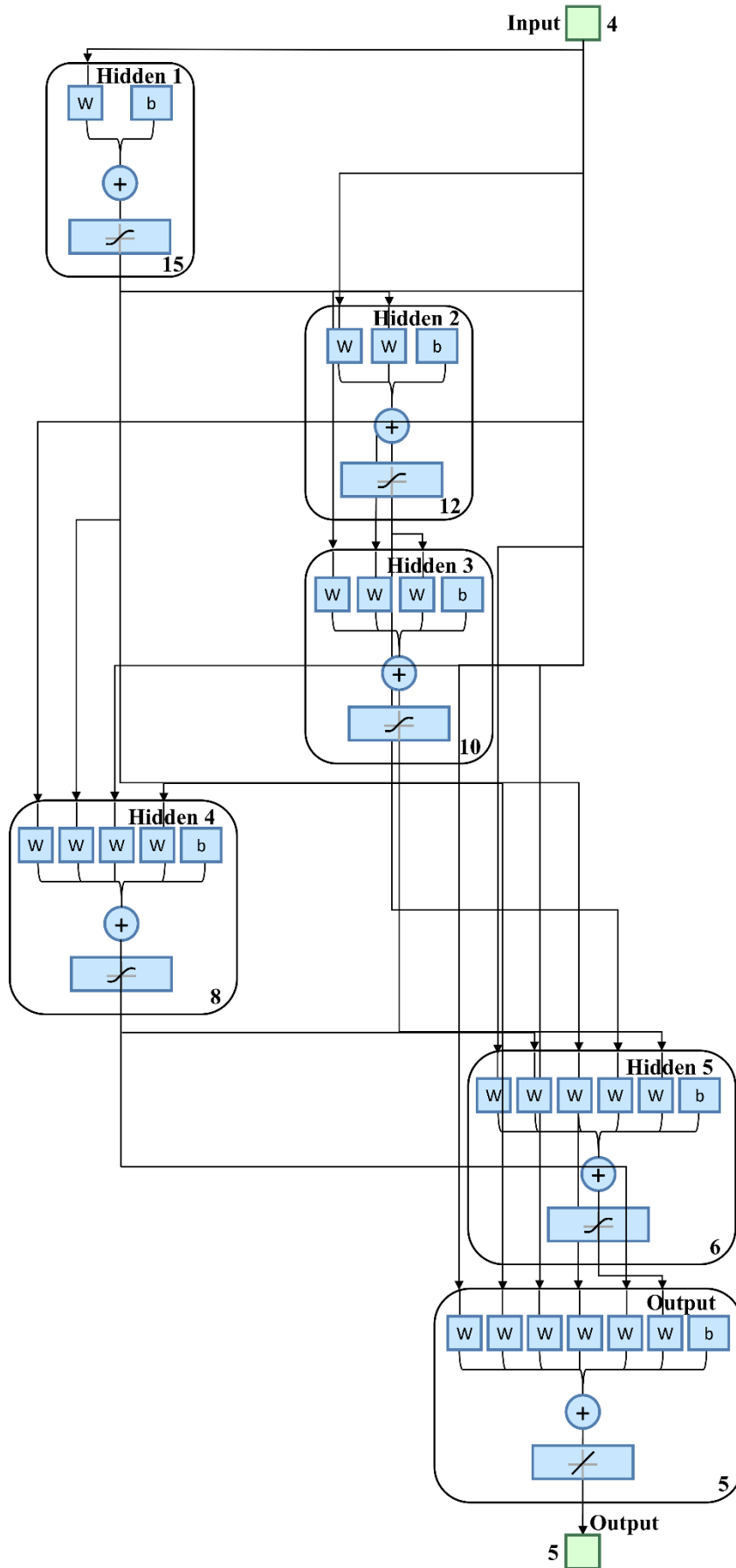


Figure 4.2. Cascade-forward ANN model for response prediction.

The training process was governed by several parameters: a maximum of 1000 epochs, performance goal of 1×10^{-7} , minimum gradient of 1×10^{-9} , and maximum of 40 validation checks. The learning rate was initialized at 0.00005 and dynamically adjusted during training, with a decrease factor of 0.15, increase factor of 1.2, and maximum allowable learning rate of 1×10^5 . The regularization was incorporated using an L2 regularization parameter of 0.0003 to prevent overfitting. The dataset was divided into training, validation, and test subsets with ratios of 0.75, 0.15, and 0.10. The weights were initialized using an improved weight initialization method to enhance convergence. Additionally, parallel processing is enabled by leveraging the available computational resources to expedite training.

To enhance the performance of the ANN and reduce the best validation performance (mean squared error), all response values were normalized prior to training. This normalization process ensures that each output lies on a consistent numerical scale, thereby preventing any single response from disproportionately influencing the loss function owing to the scale differences. It effectively reduces the high MSE values arising from magnitude mismatches, facilitates faster convergence, and improves the generalization capability of the network. Normalization was performed using the min–max scaling technique, defined as

$$x_{norm} = \frac{x - x_{min}}{x_{max} - x_{min}} \quad (4.8)$$

Where x is the original response value. x_{min} , x_{max} are the minimum and maximum values for each response across the dataset, respectively. x_{norm} is the normalized value in the range $[0,1]$.

Once the ANN model was trained, the predicted outputs (in normalized form) were converted back to their original scale for proper interpretation. This conversion was performed via a denormalization process using Equation 4.9.

$$x = x_{norm} \times (x_{max} - x_{min}) + x_{min} \quad (4.9)$$

The objective function plays a critical role in evaluating ANN performance based on the input and desired output values. It calculates a weighted sum of responses, prioritizing specific objectives through the assigned weights. Positive weights ($w = 0.2$) were used for responses to be minimized, whereas negative weights ($w = -0.2$) were assigned to

responses to be maximized. The objective function normalizes the inputs as required and provides a quantitative metric for the optimization performance, effectively guiding the algorithms toward optimal solutions.

The final optimal solution is determined using the following structured selection process:

Non-dominated solution identification: Non-dominated solutions were identified from the final population using Pareto dominance criteria. A solution was considered non-dominated if no other solution outperformed it across all objectives.

Normalization and scoring: The identified solutions were normalized on a scale of [0, 1] using the following formula:

$$\text{Normalized Value} = \frac{\text{Value} - \text{Min}}{\text{Max} - \text{Min}} \quad (4.10)$$

Compromise solution selection: For each normalized solution, the sum of all normalized objectives was calculated. The solution with the minimum sum is selected as the final optimal solution, ensuring a balanced trade-off across all objectives. This approach ensures that the selected solution is Pareto optimal, unbiased by the original objective scales, and represents the optimal compromise.

The Pareto front, representing non-dominated solutions, was visualized using a modified RadVis technique [152]. The objectives were normalized and the solutions were projected onto a 2D plane using the RadVis projection method. To enhance the interpretability of the visualization, the solutions were color coded based on the first objective value, effectively creating a pseudo-third dimension through color intensity. This 2D projection with color mapping provides an intuitive way to observe the trade-offs among objectives, where the spatial distribution of points represents the relationships between different solutions and the color gradient indicates performance in terms of the first objective. Although this approach simplifies the traditional 3D-RadVis visualization, it maintains essential information about the Pareto front structure and enables an effective analysis of the multi-objective optimization results.

Sensitivity analyses were conducted to assess the influence of input variables on the output responses. Local and global sensitivity analyses were performed using the results of the metaheuristic algorithms.

Local sensitivity analysis (one-at-a-time): This method evaluates the effect of small perturbations in individual input variables on the outputs. For each input, perturbed input vectors were generated with $\pm 1\%$ change, and the local sensitivity index was calculated using the following formula:

$$Local\ SI = \frac{Output_{plus} - Output_{minus}}{2 \times \delta \times Nominal\ Value} \quad (4.11)$$

Global sensitivity analysis (Sobol method): This variance-based approach [153] quantifies the contribution of each input variable to output variability across the input space. Using the sampled data, first-order Sobol indices were calculated for each input-output pair as follows:

$$Global\ SI = 1 - \frac{Variance_{perturbed}}{Variance_{unperturbed}} \quad (4.12)$$

Heatmaps and bar charts were used to visualize the sensitivity indices, providing insights into the relative significance of the input parameters. These visualizations facilitated comparisons and informed parameter prioritization for optimization.

4.4.1 Teaching-learning-based optimization

The TLBO algorithm, as illustrated in Figure 4.3, was implemented in two main phases: teacher and learner. Initially, a population of candidate solutions (students) was generated and their objective function values were evaluated. In the Teacher Phase, the best-performing individual was considered as the teacher, and the mean of the population was adjusted based on the teacher's influence to guide students toward better solutions. In the subsequent Learner Phase, students interacted with one another to improve their knowledge by comparing their performance and updating their solutions accordingly. This iterative process continues until the convergence criteria are met, thereby optimizing the desired coating parameters [154,155].

4.4.2 Whale optimization

As illustrated in Figure 4.4, the WOA is implemented by initializing a population of whales (candidate solutions) and evaluating their fitness. The best-performing agent is identified, and the control parameters a , A , C , l , and p are updated for each iteration. Depending on the value of p , the whales either engaged in exploitation (using encircling or spiral mechanisms) or exploration (searching for prey). The whale positions were then updated accordingly and their fitness was recalculated. If a better solution is obtained, then the current best solution is replaced. This process continues until the maximum number of iterations is reached, at which point the best solution is returned [156].

4.4.3 Red fox optimization

As shown in Figure 4.5, the RFO algorithm is implemented by initializing a population of red foxes and evaluating their fitness. The best-performing fox was identified, and the positions of other foxes were updated based on natural behaviors such as hunting, hiding, and escaping predators. These behaviors were mathematically modeled to balance exploration and exploitation during the search process. The positions of the foxes were iteratively refined and the fitness was recalculated in each iteration. The algorithm continues until the maximum number of iterations is reached, ultimately yielding the best solution [157].

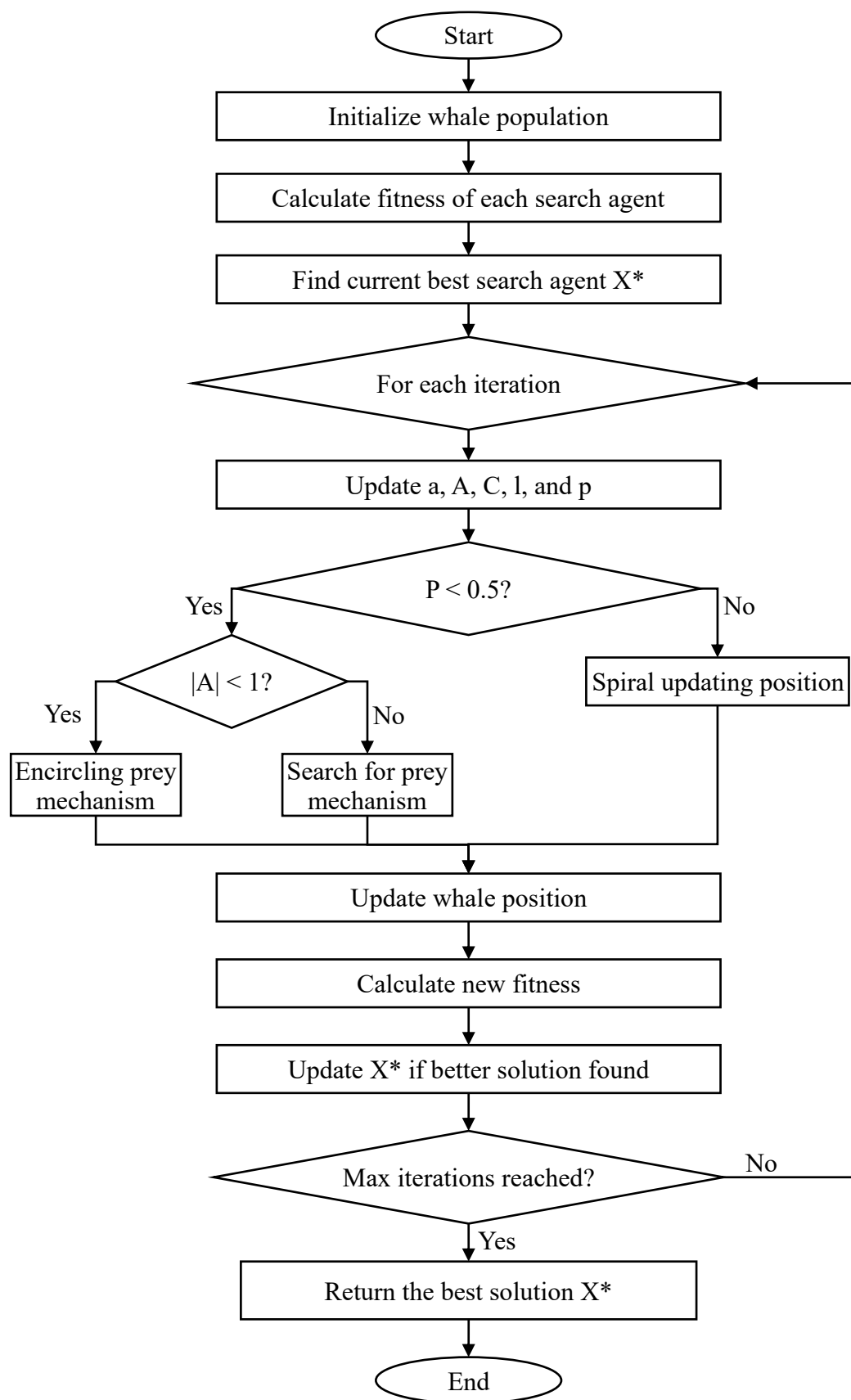


Figure 4.4. Flowchart of the whale optimization algorithm.

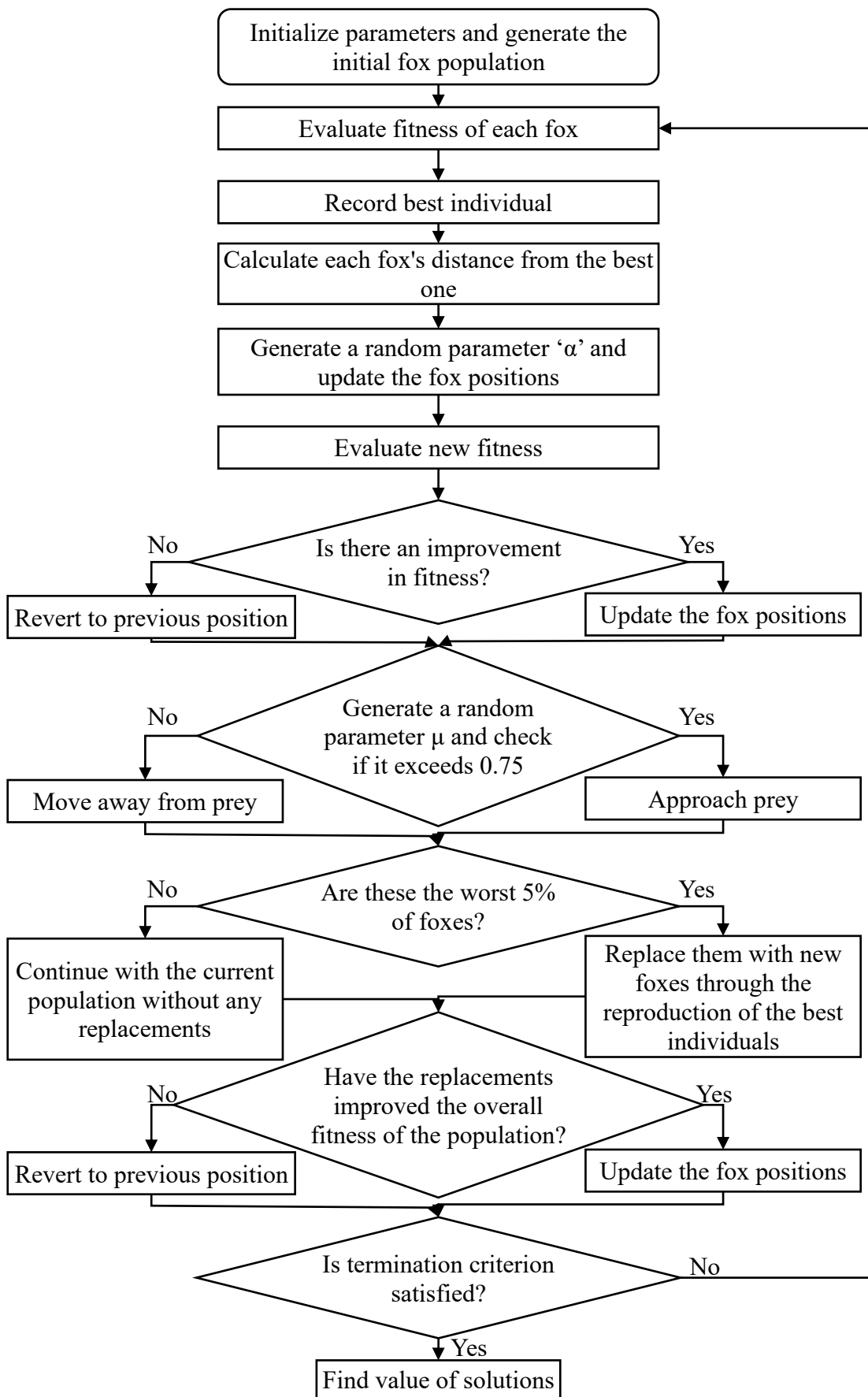


Figure 4.5. Flowchart of the red fox algorithm.

4.5 Results and discussion

The experimental results from all 30 trials, including the influencing factors and their corresponding values for the five key responses (surface roughness, microhardness, fouling weight, surface free energy, and corrosion protection efficiency), are summarized in Table 4.3. Analysis of the data revealed the performance range for each response across different experimental runs.

Surface roughness varied between 0.23 μm in the 22nd run and 1.28 μm in the 19th run, with lower values being preferred for better coating quality. Although the 22nd run achieved the best surface roughness, its performance in other key responses was not as favorable, indicating that an optimal balance was not achieved in that trial. Vickers microhardness ranged from 329.08 HV 0.05 (7th run) to 689.76 HV 0.05 (13th run), with higher values being desirable for enhanced mechanical strength. While the 13th run demonstrated the highest microhardness, it did not perform well in terms of surface roughness, fouling, or corrosion resistance, suggesting that hardness alone does not guarantee the overall coating performance. Fouling weight ranged from a minimum of 59.01 g/m^2 in the 6th run to a maximum of 110.11 g/m^2 in the 19th run. Surface free energy values spanned from 15.02 J/m^2 (6th run) to 34.15 J/m^2 (30th run). The 6th run performed notably well in both fouling resistance and surface free energy, two critical indicators for antifouling and wettability characteristics, although its performance in other responses was moderate. The corrosion protection efficiency ranged from 45.01% in the 7th run to 94.21% in the 15th run, with higher values indicating superior protection. The 15th run demonstrated excellent corrosion resistance, yet its performance in other key parameters, such as surface roughness and microhardness, was less impressive.

These observations indicate that no single run excelled in all five responses simultaneously, underscoring the need for multi-objective optimization to identify process conditions that yield balanced and high-performance coatings across all critical parameters.

4.5.1 Response surface analysis

4.5.1.1 Effect of input parameters on responses

The main effects of the input parameters on the five responses were analyzed to optimize the electroless coating process. Figure 4.6 illustrate these relationships, and the trends are discussed below.

As shown in Figure 4.6(a), the main effects plot highlights the variation in surface roughness with the input parameters. For nickel sulfate, the surface roughness decreased significantly as the concentration increased from 15 to approximately 40 g/L, after which it began to increase. This suggests that a moderate concentration of nickel sulfate achieves the optimal deposition conditions, balancing the nucleation and growth rates to produce a smoother surface. Higher concentrations likely disrupted this balance, resulting in rougher texture. Sodium hypophosphite exhibited a similar trend, where the surface roughness decreased to approximately 32.5 g/L but increased beyond this level. As a reducing agent, sodium hypophosphite enabled uniform deposition at moderate levels. However, excessive amounts may accelerate the nucleation rate, leading to uneven growth and a rougher texture because the grains lack sufficient time to coalesce smoothly [158]. In contrast, sodium tungstate exhibits a different trend. The surface roughness remained low at lower concentrations, but rose sharply beyond approximately 22.5 g/L. This is likely due to the formation of a coarse columnar growth pattern at higher concentrations, which alters the deposition kinetics and introduces internal stresses, resulting in surface irregularities [159]. Cerium sulfate also exhibited a parabolic trend, with surface roughness reaching a minimum of approximately 8 mg/L. Cerium acts as a grain refiner and promotes a uniform coating. However, at higher concentrations, the agglomeration of small CeO_2 and Ce_2O_3 crystallites increases surface roughness [160,161].

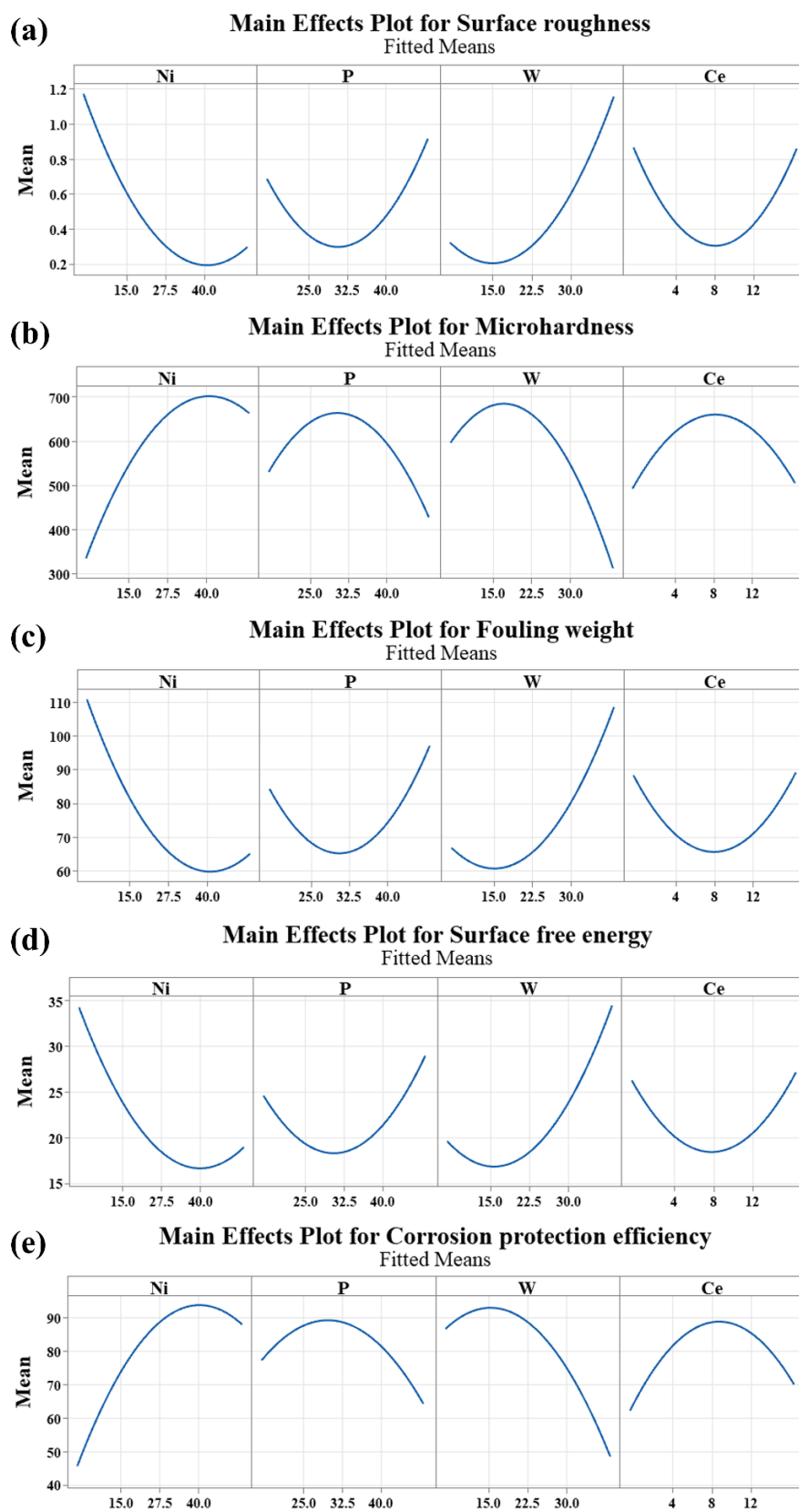


Figure 4.6. Main effect plots of Ni, P, W, and Ce on (a) surface roughness, (b) microhardness, (c) fouling weight, (d) surface free energy, and (e) protection efficiency.

Figure 4.6(b) reveals distinct trends for the input parameters that affect the microhardness. For nickel sulfate, the microhardness increased with the concentration, peaking at approximately 40 g/L before declining. This suggests an optimal Ni concentration for the formation of a dense, hard Ni matrix. Excessive nickel sulfate may lead to rapid deposition, reducing coating uniformity and creating a porous structure associated with lower hardness [162]. Sodium hypophosphite also showed an increasing–decreasing trend, with a peak within the tested range. As a reducing agent, it facilitates nickel deposition and enhances hardness; however, higher concentrations may introduce excessive phosphorus, affecting the structural integrity [158]. Sodium tungstate exhibited a peak at approximately 22.5 g/L, likely owing to the formation of a tungsten-nickel alloy phase. Beyond this concentration, the microhardness decreases as microcracks form, driven by the internal stresses caused by excess tungsten [163]. Cerium sulfate exhibited a moderate peak at approximately 8 mg/L, after which its hardness decreased. Optimal cerium levels refine the microstructure, whereas excessive amounts lead to agglomeration, reducing hardness [160,161].

Optimization of the coating parameters can minimize fouling, as illustrated in Figure 4.6(c). For nickel sulfate, fouling weight decreases steeply from 15.0 to around 27.5 g/L, then stabilizes. A uniform, smooth surface with lower surface energy reduces fouling by minimizing the adhesion of foulants [164]. Sodium hypophosphite followed a similar trend, with a minimum fouling weight of approximately 32.5 g/L. Optimal concentrations promote a compact structure, whereas higher concentrations may result in excessive phosphorus, rougher surfaces, or structural defects, thereby increasing fouling [158]. Conversely, sodium tungstate exhibits an increase in fouling weight with increasing concentration, which is attributed to its reduced thermal conductivity and higher surface energy. These factors enhance adhesion and create an environment conducive to fouling [165]. Cerium sulfate displayed a parabolic trend, with minimum fouling at approximately 8 mg/L. Excessive cerium leads to agglomeration, disruption of deposition, and increasing fouling [121].

Figure 4.6(d) shows that the SFE declined steeply with increasing nickel sulfate concentration, reaching a minimum at 27.5 g/L. This decrease is linked to smoother and denser surfaces with fewer molecular interactions [166]. Sodium hypophosphite exhibited a similar trend, with SFE stabilizing beyond 32.5 g/L. Excess phosphorus increases the SFE owing to changes in surface composition and structure [167]. In contrast, sodium tungstate

increased SFE with increasing concentration. Higher tungsten levels alter the surface morphology and chemical composition by introducing polar functional groups that enhance SFE [168]. Cerium sulfate followed a parabolic trend with a minimum SFE of 8 mg/L. Beyond this level, cerium agglomeration leads to rougher surfaces and higher SFE.

Figure 4.6(e) shows that nickel sulfate increased the efficiency to 40 g/L, after which it declined. Higher nickel sulfate levels enhance the deposition uniformity, but excessive amounts cause spontaneous decomposition and nickel hydroxide formation, compromising the coating quality [73]. Sodium hypophosphite exhibited a similar parabolic trend, with the efficiency peaking at 32.5 g/L. Excessive phosphorus introduces microstructural defects and reduces corrosion resistance [169]. Sodium tungstate exhibited maximum efficiency near 15 g/L. Higher concentrations result in defective passive films, thereby decreasing protection [170]. Cerium sulfate optimized the corrosion resistance to approximately 8 mg/L by enhancing the compactness and thickness of the coating. Excessive cerium introduces structural irregularities and reduces efficiency [143,171].

4.5.1.2 ANOVA and regression analysis

The significance of the input parameters for each response was evaluated using ANOVA considering their linear, quadratic (square terms), and interaction effects. The ANOVA results for the surface roughness, microhardness, fouling weight, surface free energy, and corrosion protection efficiency are listed in Tables 4.4, 4.5, 4.6, 4.7, and 4.8, respectively. Significant factors were identified by p-values less than 0.05, which are marked with an asterisk (*). The four input parameters, nickel sulfate, sodium hypophosphite, sodium tungstate, and cerium sulfate, are denoted in their coded forms as Ni, P, W, and Ce, respectively, in certain paragraphs of this chapter.

The ANOVA for surface roughness indicated that Ni and W were the most influential factors, with their linear and quadratic terms exhibiting high statistical significance ($p = 0.000$). This demonstrates the critical role of these elements in determining the surface roughness. Furthermore, the interaction between Ni and Ce was significant, suggesting a synergistic effect between these factors on surface roughness. The model provided an excellent fit, as evidenced by an R^2 value of 98.64%, indicating its ability to accurately capture the variability in surface roughness. The non-significant lack of fit supports the adequacy of the model in explaining the observed variation.

g/L sodium hypophosphite produces a smooth, low-energy surface. The addition of 15 g/L of sodium tungstate, as shown in Figure 4.11(b), further reduced the surface energy through enhanced compactness. Figure 4.11(c) shows that the interaction between 40 g/L nickel sulfate and 8 mg/L cerium sulfate refined the surface, thereby minimizing irregularities. Figure 4.11(d) shows that 30 g/L of sodium hypophosphite and 15 g/L of sodium tungstate yielded a dense coating with reduced surface energy. The combination of 30 g/L sodium hypophosphite and 8 mg/L cerium sulfate achieved the lowest surface free energy through the grain refining effect of cerium, as shown in Figure 4.11(e). Finally, Figure 4.11(f) shows that 15 g/L sodium tungstate and 8 mg/L cerium sulfate produced a uniform low-energy surface.

The surface plots for corrosion protection efficiency (Figure 4.12(a-f)) show the conditions for achieving the maximum resistance to corrosion. As shown in Figure 4.12(a), the combination of 40 g/L of nickel sulfate and 30 g/L of sodium hypophosphite formed a dense protective layer. Figure 4.12(b) shows that 15 g/L sodium tungstate with 40 g/L nickel sulfate enhanced corrosion resistance through tungsten incorporation. The addition of 8 mg/L cerium sulfate, as shown in Figure 4.12(c), improves the protection efficiency by refining the microstructure. Figure 4.12(d) indicates that the co-deposition of 30 g/L sodium hypophosphite and 15 g/L sodium tungstate enhances the robustness of the coating. The interaction of 30 g/L sodium hypophosphite and 8 mg/L cerium sulfate, as shown in Figure 4.12(e), achieved the maximum protection efficiency through the grain refining effect of cerium. Finally, Figure 4.12(f) demonstrates that 15 g/L sodium tungstate and 8 mg/L cerium sulfate synergistically enhance surface stability and maximize corrosion protection.

