

Optimization of antioxidant extraction through artificial neural network

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IECAN
2025
Conference

The 2nd International Electronic
Conference on Antioxidants

07–09 April 2025 | Online



Universidade de Vigo



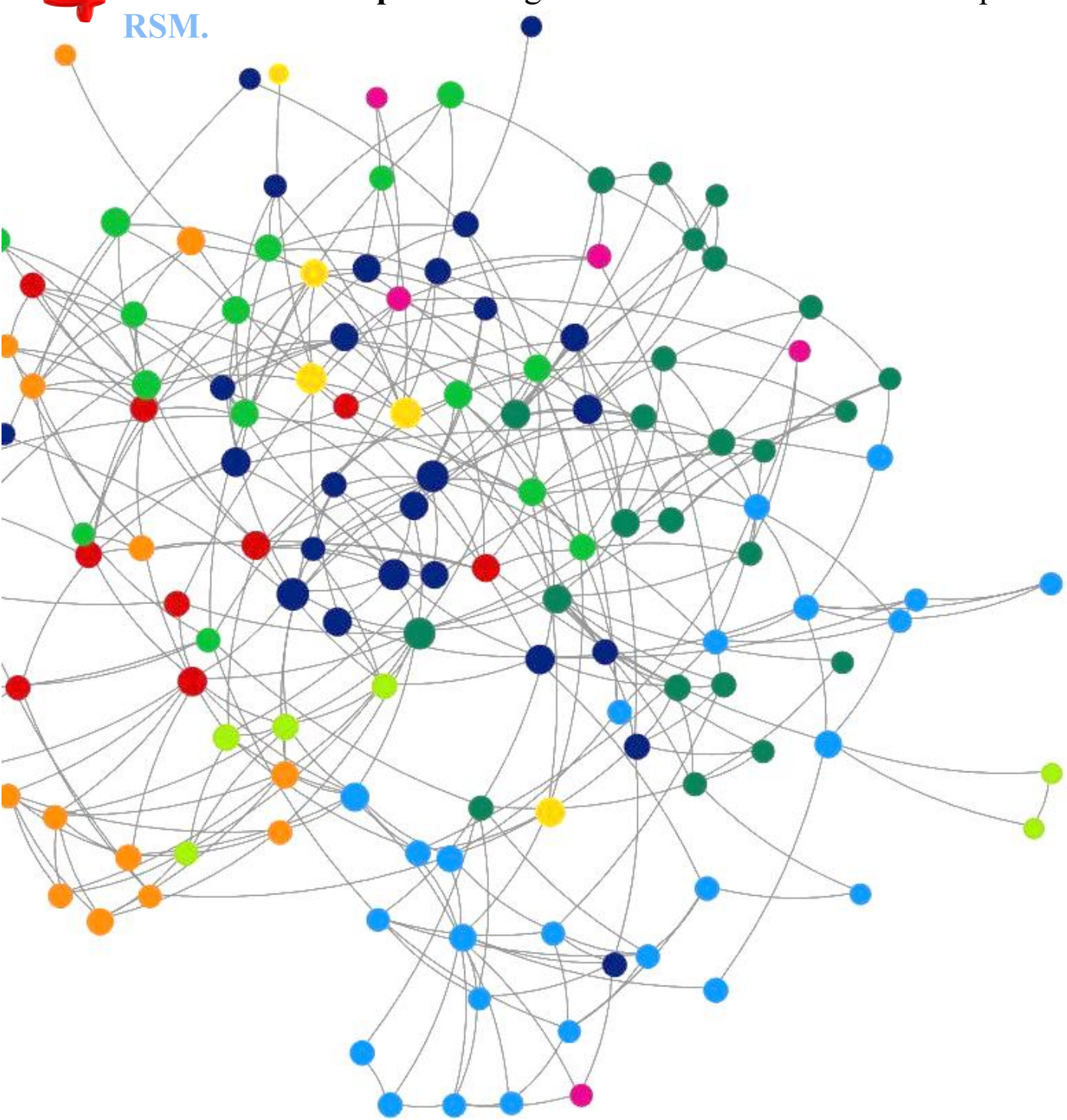
Introduction

The increasing demand in modern society for natural sources of bioactive compounds with potential applications in preventive medicine has driven the development of **innovative approaches** to optimize extraction processes. While **Response Surface Methodology (RSM)** has been extensively utilized for **prediction** and **optimization** in extraction studies, **Artificial Neural Networks (ANNs)** offer a **powerful** alternative by enabling the **development** of **non-linear computational models** based on **Deep Learning (DL)**.

These models are capable of **self-learning** and **training** to address **practical challenges**, without relying on **pre-defined mathematical equations**, by mimicking the **structure** and **functionality** of **biological neural networks**. The application of **ANNs** offers numerous **advantages**, including the ability to **interpret complex datasets**, and **scale results through optimization**. Furthermore, **ANNs** allow handling large datasets and generalizing across systems, **all without the limitations of predefined models or specific experimental designs**.

Objectives

- 1 Explore the impact of **ANNs** on enhancing sustainable and green extraction technologies.
- 2 Explore the **applicability of ANNs** in **designing industrial processes for incorporating natural antioxidants**.
- 3 Compile real studies on **ANN** application in optimizing antioxidant extraction from **natural matrices**
- 4 Offer **novel insights** into improving **extraction efficiency** and **resource consumption** using **ANNs** over traditional techniques like **RSM**.



Acknowledgments

The research leading to these results was supported by the Xunta de Galicia with the predoctoral grant of P. Barciela (ED481A-2024-230). The authors are also grateful to the National Funding by FCT, Foundation for Science and Technology, through the research grant of A.O.S. Jorge (2023.00981.BD).

Final remarks and outstanding questions

ANNs improve antioxidant extraction by efficiently managing complexity, predictive accuracy and resource utilization. However, challenges such as **data quality**, **model interpretability**, and **scalability of ANN models** for large-scale industrial use remain critical obstacles.

Results

What exactly are ANNs? Are they useful?

ANNs are **AI** models inspired by the human **brain**, consisting of **interconnected** **neurons**. They are used to solve **complex problems** such as **prediction**, **classification**, and **regression** by learning from data.

ANNs in antioxidant extraction processes



Advanced modeling: **ANNs** surpass **RSM** by effectively handling complex, **non-linear relationships** in antioxidant extraction optimization.



Adaptive learning: Multilayer Perceptron-based **ANNs** process data, **recognize patterns**, and **optimize parameters** for improved efficiency.



Improved accuracy: **GAs** refine **ANN** models to increase **accuracy**, reduce **resource consumption**, and **promote sustainability**.

Optimized antioxidant extraction workflow

- 1 **Parameter Selection:**
 - **Solvent mixture components:** and .
 - **Process parameters:** Extraction , solid-to-liquid (S/M) ratio, and .
- 2 **Experimental Design:**
 - **RSM:** Optimize extraction , , and to enhance antioxidant yield.
 - **Mixture design:** Optimize solvent mixtures (e.g.,) for extraction based on properties.
 - **Optimal Mixture Process Design (OMP):** Combine **RSM** and **Mixture Design** to optimize both conditions and composition, using **D-optimality** to select the design for a of experimental trials.
- 3 **Mathematical Model:**
 - **Mixture model:** Combines non-linear effects and second-order process variables effects for .
 - **Process model:** Evaluates main effects, , and second-order effects of process variables like and S/M ratio.
- 4 **Modeling with ANNs:**
 - **ANNs** to predict antioxidant yields by analyzing complex . The model uses **inputs** with **outputs** representing antioxidant activity (yield, TPC, DPPH IC₅₀, TAC, FRAP EC₅₀).
 - Validation through **k-fold cross-validation (K=4)** with of the data used for and for validation.
 - Evaluate significance using **R²**, Root Mean Square Error (RMSE), and Mean Absolute Difference (MAD).
 - **Generic Algorithms (GAs) optimization:** Enhances ANN optimization via **selection**, **crossover**, and .
- 5 **Evaluation and Validation:**
 - Validate ANN predictions by the optimal conditions to ensure antioxidant from samples.

Applications of ANNs in antioxidant extraction

Luangsakul et al., (2025) optimized the **ultrasound-assisted extraction (UAE)** of flavonoids from *Gnaphalium affine* D. Don (chewcut) using the **Box-Behnken design**. They compared **ANN** and **RSM**, with **ANN** providing more accurate predictions. Optimal conditions (**65.92°C**, **58.22% ultrasound power**, **37.95 min**) achieved **high yields of antioxidants**, suggesting the potential of chewcut as a **natural, functional ingredient** for health-oriented foods.

Harkat-Madouri, et al., (2025) applied RSM and ANNs to optimize **microwave-assisted extraction (MAE)** of total phenolic compounds from almond skins. **ANNs** ($R^2 = 0.99$) outperformed RSM ($R^2 = 0.97$) in predicting optimal conditions: **562 W microwave power**, **30 s extraction time**, and **53% EtOH**, yielding **560.79 mg GAEs/100g**. The extract showed **strong antioxidant and antihyperglycemic activities**, including **α -amylase inhibition (IC₅₀ = 27.87 μ g/mL)**.

- 1 How to scale **ANNs** for industrial applications without compromising **yield** or **cost**?
- 2 Can **DL** combined with traditional methods **outperform resource optimizing**?
- 3 How well can **ANNs** predictions be **generalized** across different food matrices?
- 4 Can **AI-powered optimization** help energy **efficiency** and sustainable mining at scale?