



Neural Networks and the Future of Farming: How Artificial Neural Networks Are Transforming Agricultural Decision-Making

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The 21st century is witnessing several forces that are changing the way agriculture is practiced. Some of these forces include climate change, dwindling natural resources, and an ever-increasing number of hungry mouths to feed. While classical statistical analysis has worked for many years, it has some limitations when it comes to the complex, nonlinear, and interrelated aspects of real-world agriculture. It is in this setting that the Artificial Neural Network (ANN), which models this human brain function, is rising to prominence as a useful tool for agriculture analysis. This article describes ANN simply and clearly, explaining its growing usage in agriculture analysis and decision making, using real-world illustrations of its usage in predicting crop yield, measuring climate change effects, and resource management for agriculture, as a complement to classical statistical analysis, providing a powerful tool for agriculture statisticians to analyse complex agriculture of the modern world.

Keywords: Artificial Neural Network, Crop Yield Prediction, Climate Variables, Machine Learning, Agricultural Statistics, Decision Support Systems, AI for Agriculture, AI Farming, Smart Farming.

Introduction

Agriculture is now at a special crossroads because traditional methods are converging with technological innovations. Today's farmers and researchers are no longer challenged with isolated issues such as precipitation and fertilizer, separately. Precipitation doesn't exist independently of temperature; soil health is no longer a separate response based on management; markets are no longer a separate factor influencing agricultural decisions either. They are all correlated. For a considerable period, regression analysis and Analysis of Variance (ANOVA) were the staple tools for agricultural research. These tools are still effective models. Yet, these models are based on a number of hypotheses that agricultural data does not comply with. Variety, uncertainty, and complexity are regular occurrences in agriculture. These differences prompted an interest in Artificial Neural Networks (ANNs), a data-focused alternative that identifies patterns from data without trying to fit data into a formula (Haykin, 2009).

What Exactly Is Artificial Neural Network?

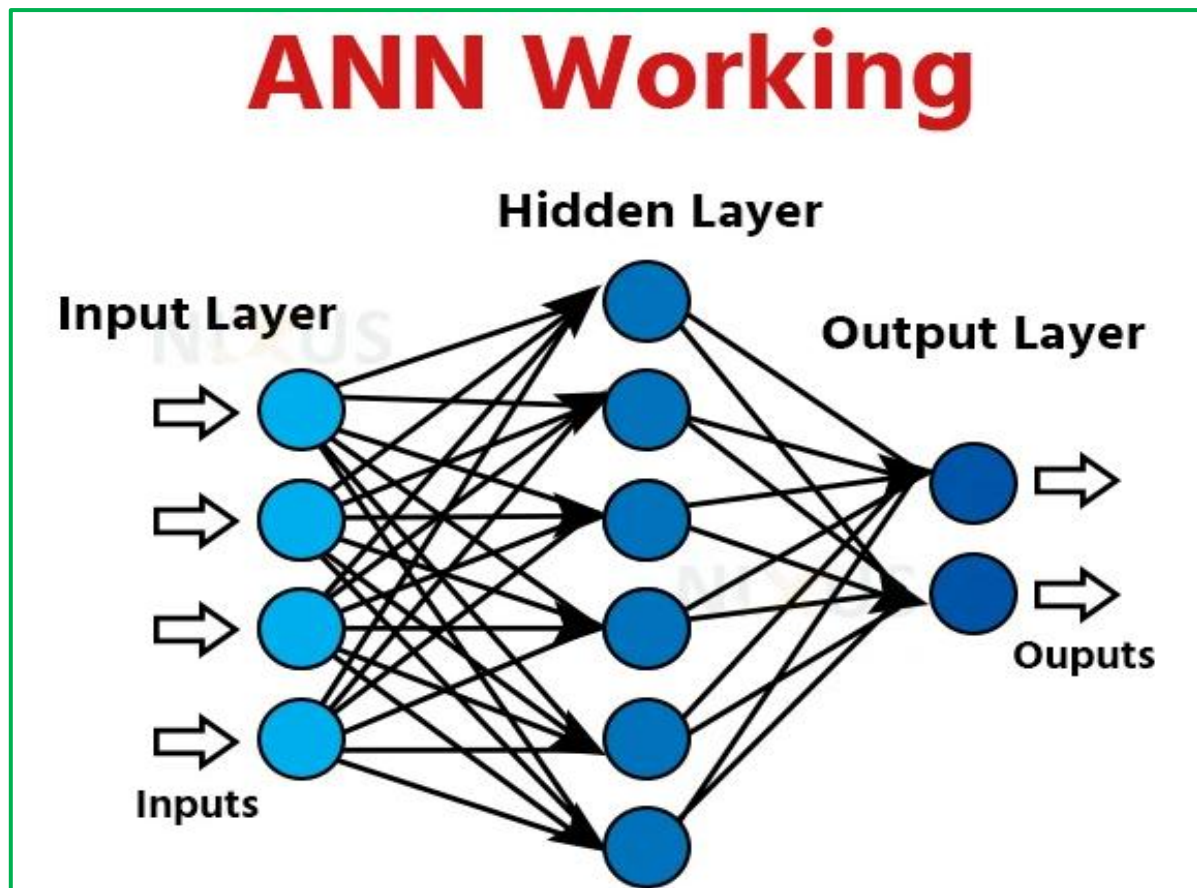
We know a human nervous system is a complex network of neurons. Artificial Neural Network is a computational model that mirrors the human nervous system. ANN consists of artificial neurons that function together to process data and learn from it.

Input Layer: This is where data on rainfall, temperature, humidity, or soil nutrients is provided.

Hidden Layer(s): This layer is responsible for internal processing where it determines the relationship between the variables.

Output Layer: This is layer gives the final output after the data is processed as per requirements. For example, the estimated crop output.

The key factor that differentiates ANN from traditional models is that there is no requirement of defining a precise relationship between inputs and outputs, as this is something that the model executes by itself during training, and getting optimized as it is fed more data (Goodfellow et al., 2016).



(Source: <https://nixustechnologies.com/artificial-neural-network-in-machine-learning/>
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Why ANN Is Gaining Importance in Agriculture?

Agricultural systems are nonlinear. For instance, a small rise in temperature may result in a negligible effect on the yield of crops when the land is well-irrigated, while the same effect may cause a drastic reduction in yield when it comes to drought conditions. Such phenomena are difficult to interpret using linear models.

ANN tackles this problem by automatically adapting it to complexity. The main advantages of ANN are:

- Capacity for modelling non-linear and complex connections.
- Little reliance upon strict statistical assumptions.
- High predictive accuracy.
- Managing large and multivariate data efficiently.

Since, variation and uncertainty are inherent in agricultural studies, using ANN suits the purpose of conducting the studies well.

Applications of ANN in Agriculture

1. Crop Yield Prediction: Models of ANN employ climatic variables like rainfall, temperature, humidity, wind speed, and sunshine hours, which prove more accurate in predicting the produce of crops than many other techniques. This increases the capacity of farmers, planners, and policymakers to make appropriate decisions regarding the use of these inputs, storage, and food security planning (Ramesh & Vardhan, 2015).

2. Climate Impact Assessment: Climate change has different effects on agriculture. There is no predictability in this regard. ANN is useful for establishing complex connections between climate and crops. Scientists can identify risks and accordingly design adaptation measures.

3. Soil and Water Management: Water and soil resource management is very important in the context of sustainable agriculture practices. ANN has been increasingly used for optimal management of irrigation systems and soil fertility estimation, especially when the resources are not abundantly available (Shankar *et al.*, 2020).

4. Pest & Disease Forecasting: To explain with an example, by studying the weather patterns and crop conditions in the past, the ANN can predict the infestation of insects and occurrence of crop diseases. With that, farmers can prepare preventive strategies in place of remedial actions.

Is ANN a ‘Black Box’?

One of the criticisms of the ANN is that it acts like a “black box,” producing predictions with a high degree of accuracy but providing no explanations for their results. Although it was a problem in the past, current advances have brought about improvements in interpretability. **Garson’s algorithm, Olden’s technique, SHAP values, and Partial Dependence Plots** help to interpret how a specific variable affects the final result (Olden & Jackson, 2002). Consequently, ANN has nowadays been able to provide not only predictions, but also scientific insights.

Artificial Neurons and the Future of Agricultural Statistics

ANN does not substitute classical analysis in deeds undertaken by students or researchers in agricultural statistics. On the contrary, ANN extends the reach of the tools. There is still the need to have good basics in understanding data structure, experimental design, or even statistical skills. ANN is built on the same premise. With agricultural research continuing to yield bigger and more intricate databases, the skill to merge statistical knowledge with machine learning algorithms will be at the forefront of the future of agricultural research and education.

Conclusion

Artificial Neural Networks are a significant paradigm change in the analysis and interpretation of agricultural data. Instead of learning from experiences and reducing complexities in data, ANN makes accurate predictions and interpretations that are essential in decision-making. In this era characterized by uncertainties in climate change and issues relating to food security, ANN is a vital and timely approach that should be adopted in agricultural studies. Being an agricultural statistician, learning and applying ANN is not only an academic necessity in current studies—it is also a move towards developing smart agricultural systems.

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