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TensorFlow Playground: Reflective Journal

Introduction

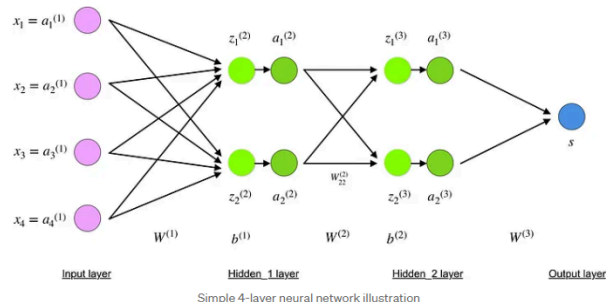
In this reflective journal, I describe my experience with an interactive web-based tool, TensorFlow Playground, for having a better knowledge on how artificial neural networks (ANNs) work. The goal is to overview ANNs, share my reflections, highlight the key findings, and the strategies deployed to navigate through the eventual challenges.

Artificial Neural Networks

Artificial neural networks, designed to mimic the brain's pattern and decision-making capabilities, are a foundational element of artificial intelligence. They are structured with layers, which interpret information through the execution of mathematical operations on weighted inputs and biases, through the activation function.

According to Kızrak, a Ph.D. AI Specialist, “Activation functions play a key role in neural networks, so it is essential to understand the advantages and disadvantages to achieve better performance” and “We need the activation function to introduce nonlinear real-world properties to artificial neural networks.”

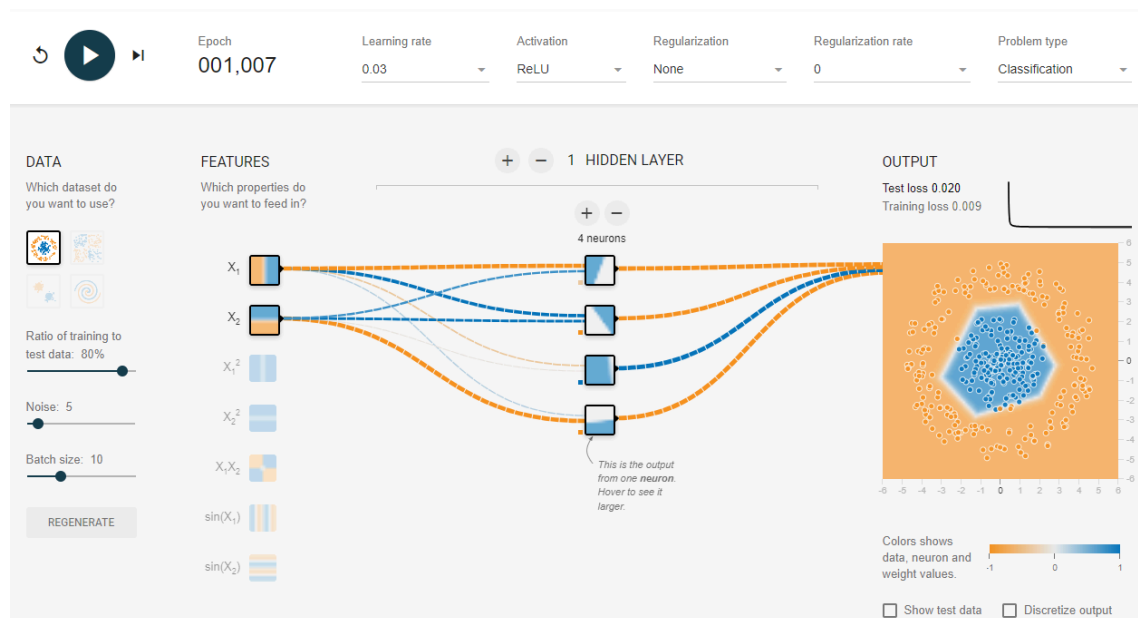
The architecture of a neural network encompasses an input layer to receive data, at least two hidden layers for processing, and an output layer to deliver predictions. Its capacity to modify the weights of connections in response to prediction errors is a method known as backpropagation, confirms Kostadinov, “backpropagation aims to minimize the cost function by adjusting network’s weights and biases”. This adaptive process is the



core of neural network learning and it enhances the precision of neural networks over time.

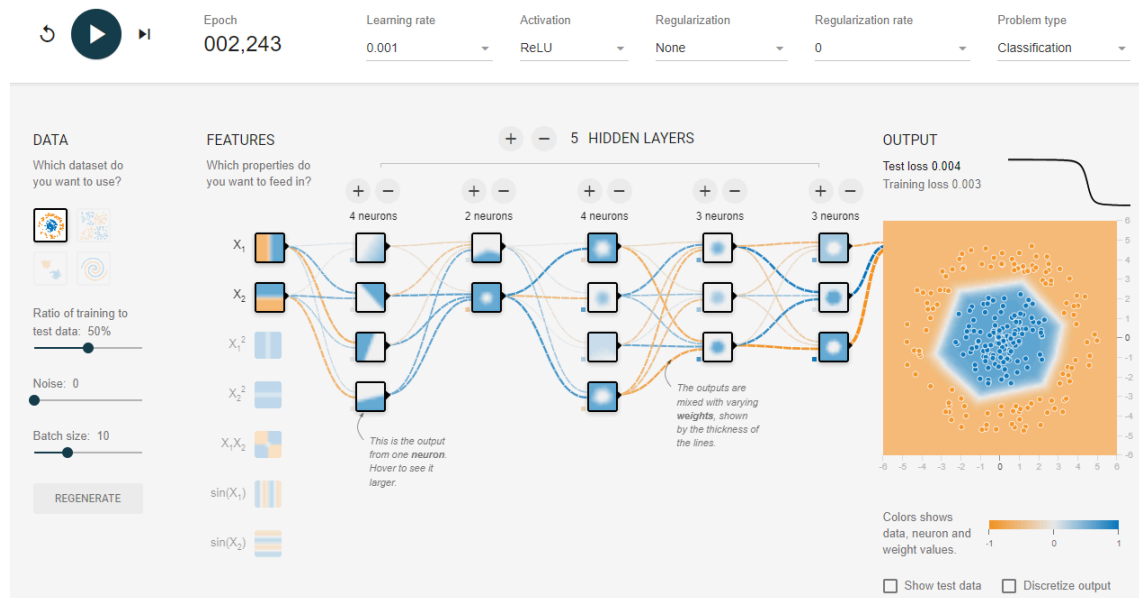
Personal Reflection

By engaging in a series of tasks with TensorFlow Playground, I have acquired key insights into artificial neural networks behavior and performance. Experimenting with different activation functions like ReLU and sigmoid highlighted their difference on network efficiency, with ReLU enabling quicker learning of complex patterns. Adjusting the number of neurons and layers affects network capacity and complexity, shed light on the importance of strategic configuration to prevent overfitting.



Tweaking the learning rate illustrated its critical role in determining convergence speed and accuracy, where too high a rate could lead to instability, and too low a rate

might slow down learning.



Additionally, introducing noise into the data demonstrated the network's capacity for generalization, with moderate noise improving robustness but excessive noise degrading performance. Last but not least, the Playground's various datasets underscored the adaptability of necessitating neural networks to different complexities such as classification or regression problems.

Improvements and Learning

One key realization was the necessity of balancing model complexity with computational efficiency to avoid overfitting, while still capturing the nuances of the data. The experimentation process also underscored the importance of parameter tuning, especially in terms of activation functions, learning rates, and layer configurations, to optimize model performance.

Conclusion

My exploration of TensorFlow Playground has been an enlightening experience, offering a practical lens through which to view the complex workings of artificial neural networks. This interactive tool has enhanced my comprehension of neural network architecture, learning mechanisms, and enabled me to visualize the impact of various parameters on model performance.

Works Cited

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