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WHITING SCHOOL  
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# Introduction to Neural Networks

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605-447.71/625-438.71

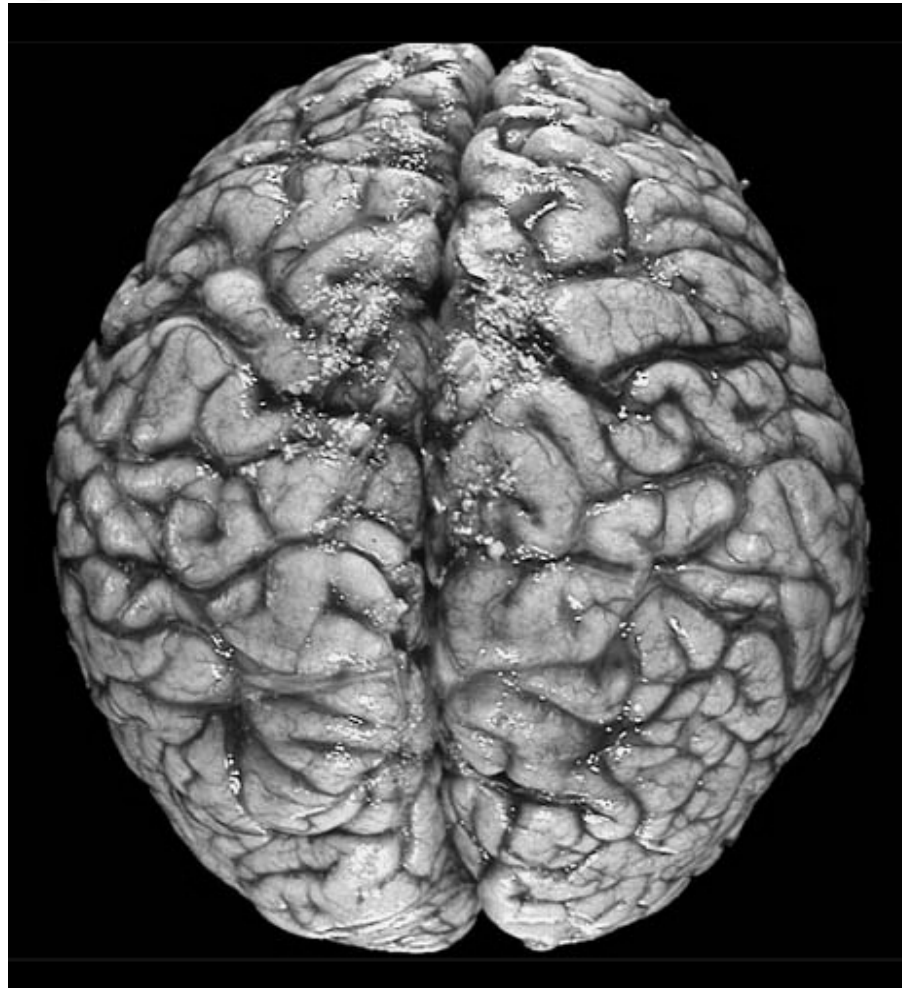
Dr. Mark Fleischer

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**Module 1.3: Modeling Considerations**



# My Brain, Your Brain



# In this sub-module...

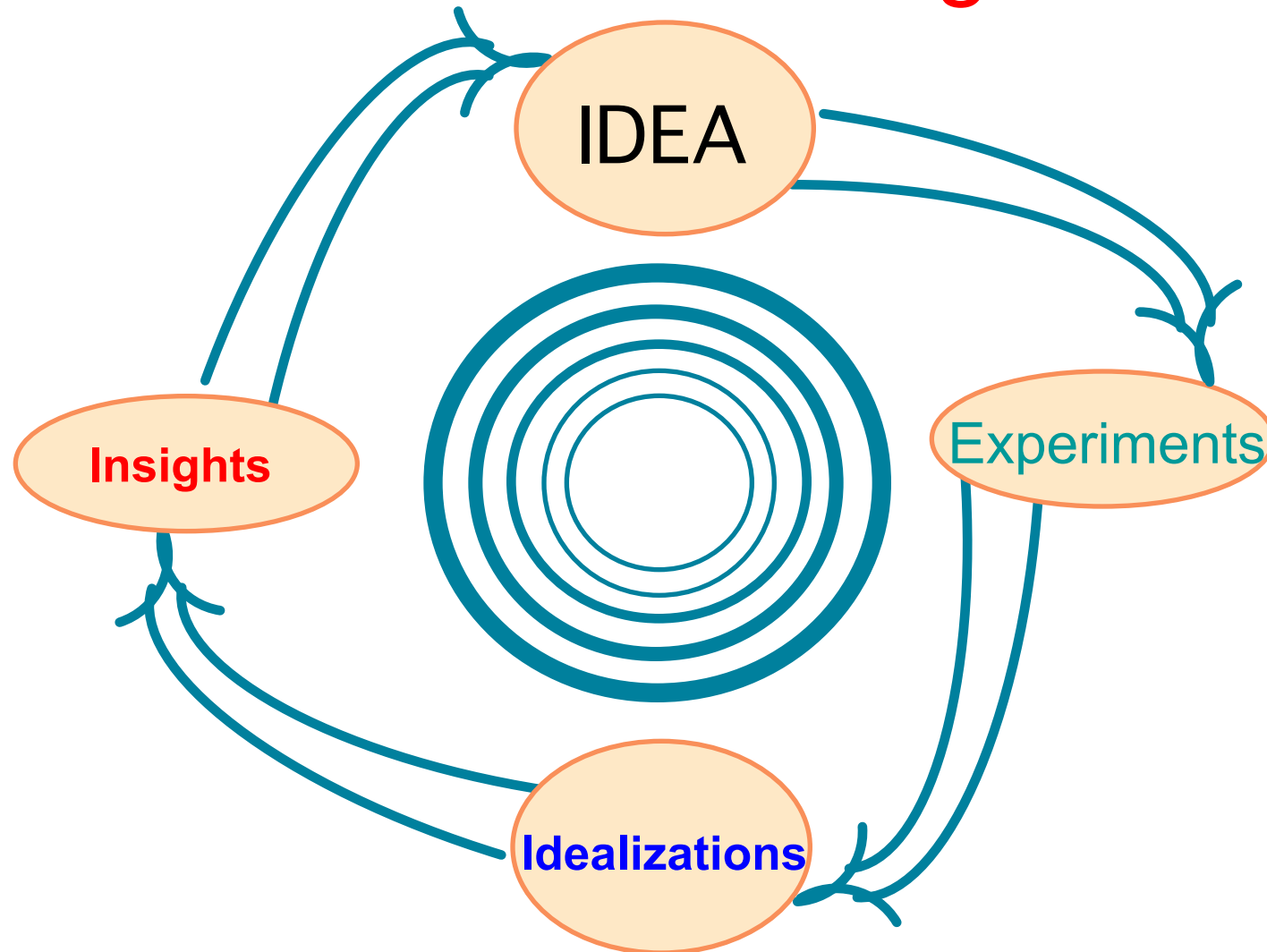
- We will cover some of the issues surrounding the art and science of modeling.
- These issues involve a tradeoff between detail and simplicity.
  - Want enough detail to capture important phenomena.
  - Want to keep things simple enough so that we can analyze our model and gain insight.

# Modeling Reality

- Can't model all details. Too difficult and misses the mark.
- Don't want it too simple...also misses the mark.
- Have to strike a balance so that some interesting phenomena can be illuminated and possibly analyzed and studied.
- Capture the essential and interesting features.

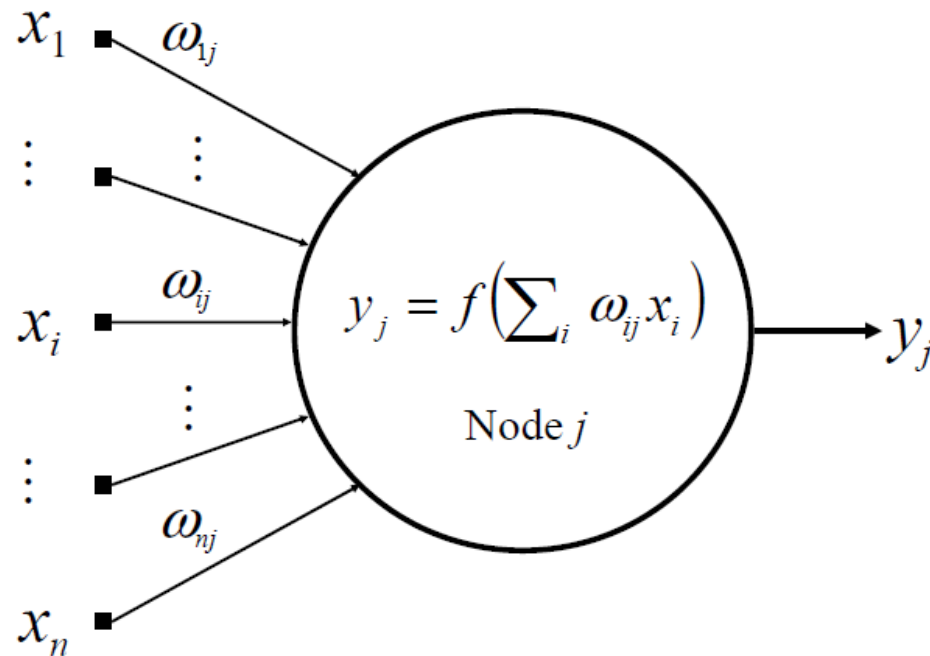


# The Vortex of Progress



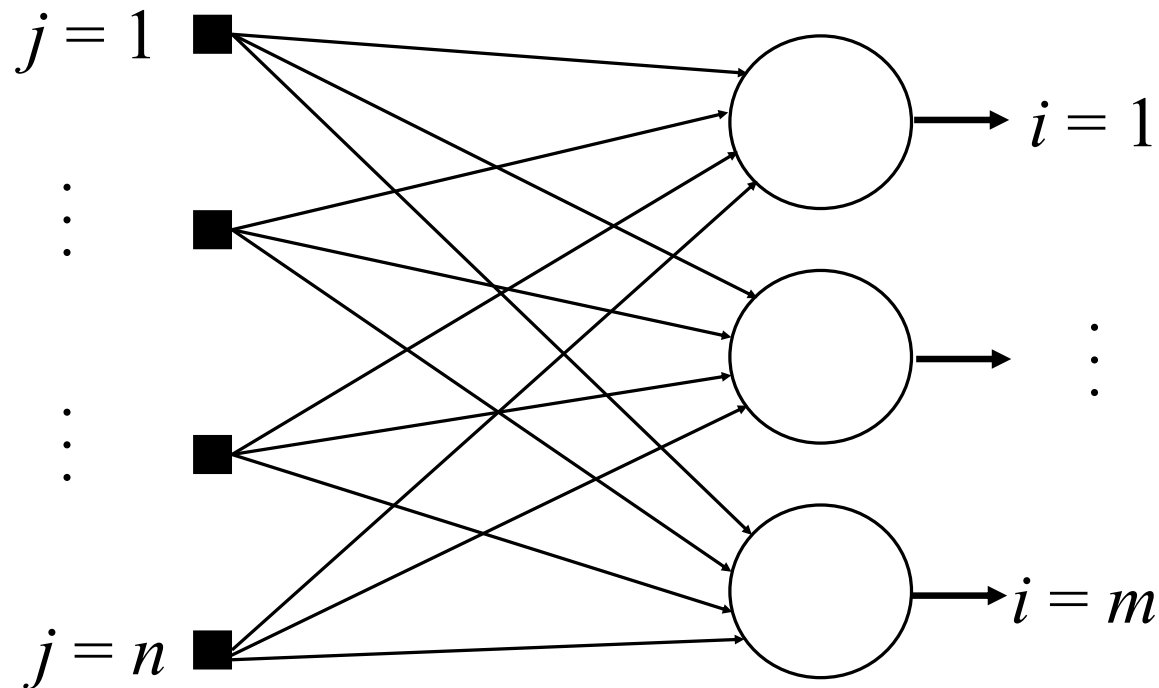


# The Perceptron





# The Multi-perceptron





# The Activity Function

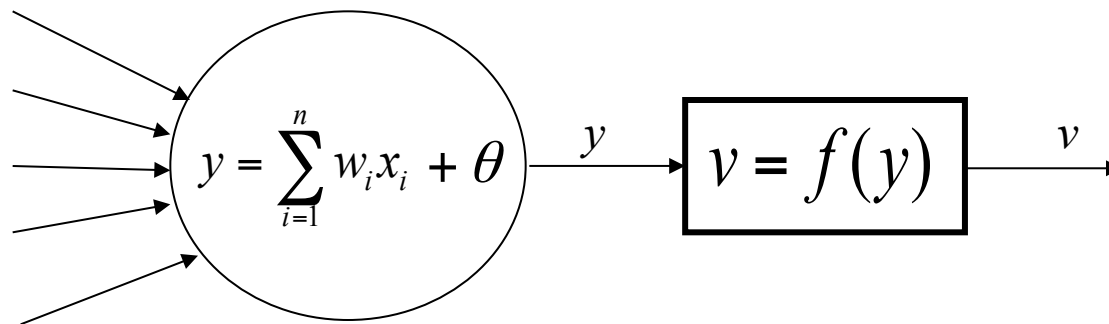
- The output of the Perceptron  $y$  is given by the equation:

$$y = \sum_{i=1}^n w_i x_i + \theta$$

- This Activity Function is sometimes referred to as a *linear basis function*.

# The Activation Function

- The Activation Function provides a one to one mapping between the Activity Function as input and some value for the output.
- It attempts to provide further flexibility in modeling biological neurons.

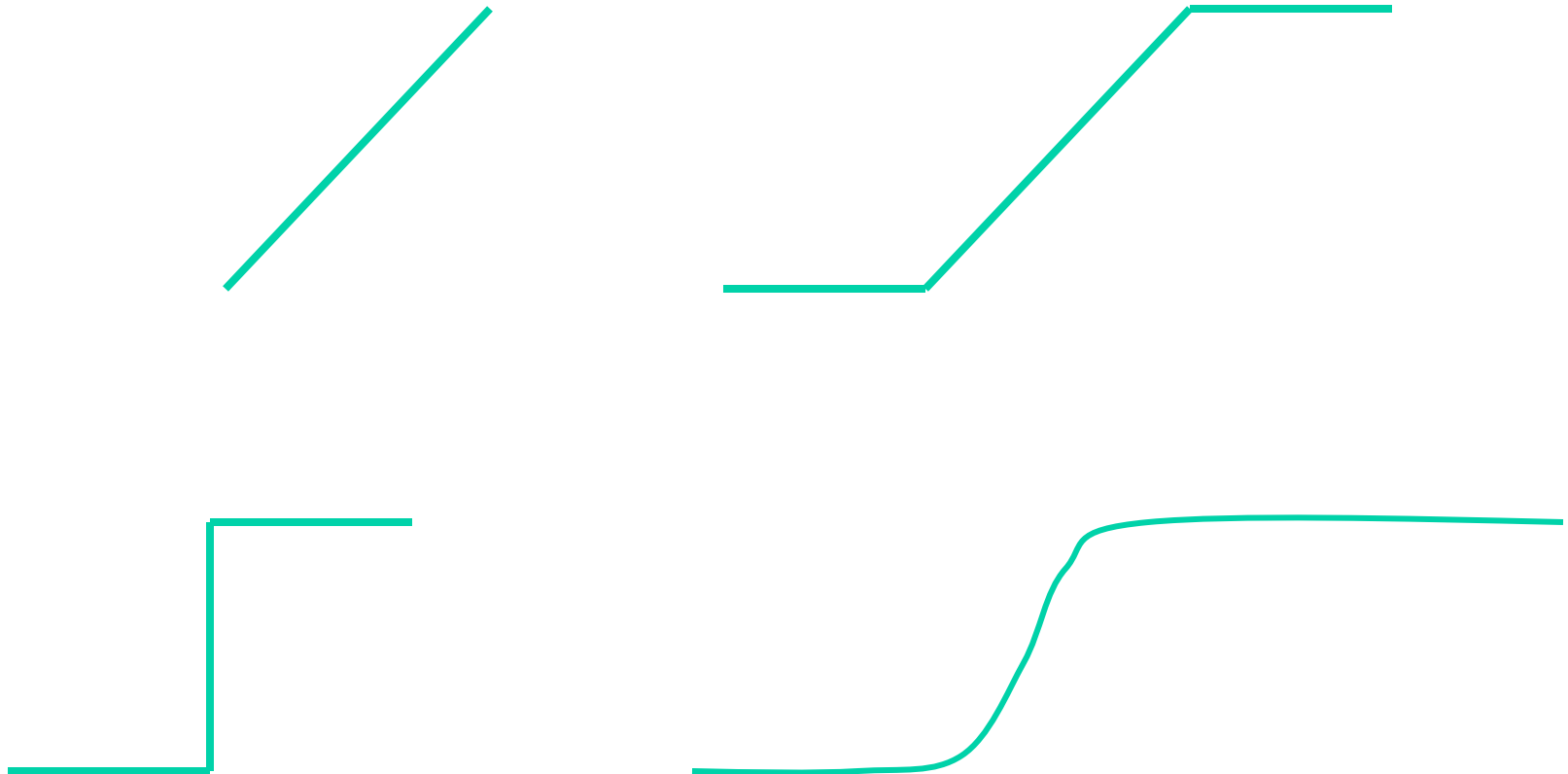


# Features of the Perceptron

- Weighted inputs --- correspond to synaptic inputs.
  - Can involve both **positive and negative values**.
  - Provides an analogy to **excitation and inhibition**.
- Activation function --- corresponds to action potential.
  - We have great flexibility in choosing the activation function.
  - We choose one that satisfies our needs.



# Activation Functions





# An Important Activation Function

## The Sigmoid Function

$$v = \frac{1}{1 + e^{-y}} = \frac{1}{1 + e^{-\left(\sum_{i=1}^n w_i x_i + \theta\right)}}$$