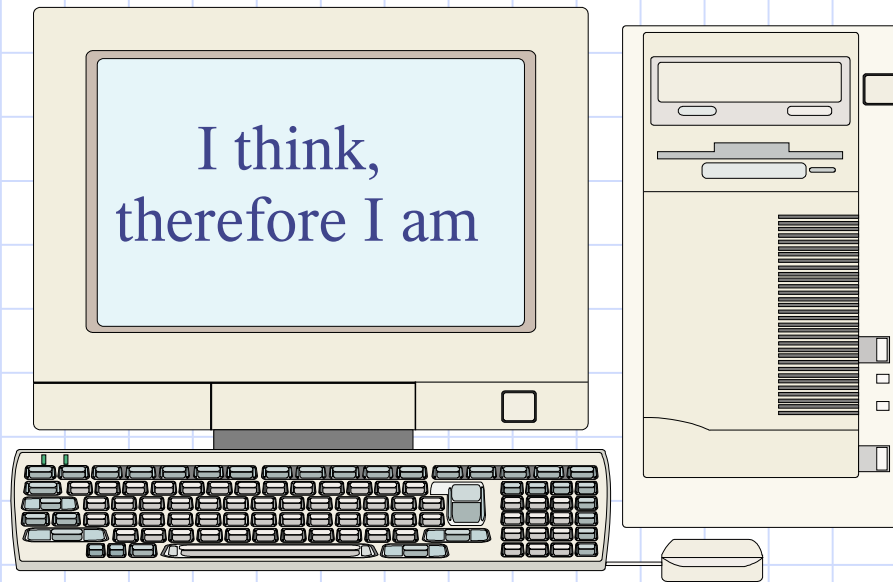


Neural Networks

A Statistical View

Brad Morantz PhD

The Future

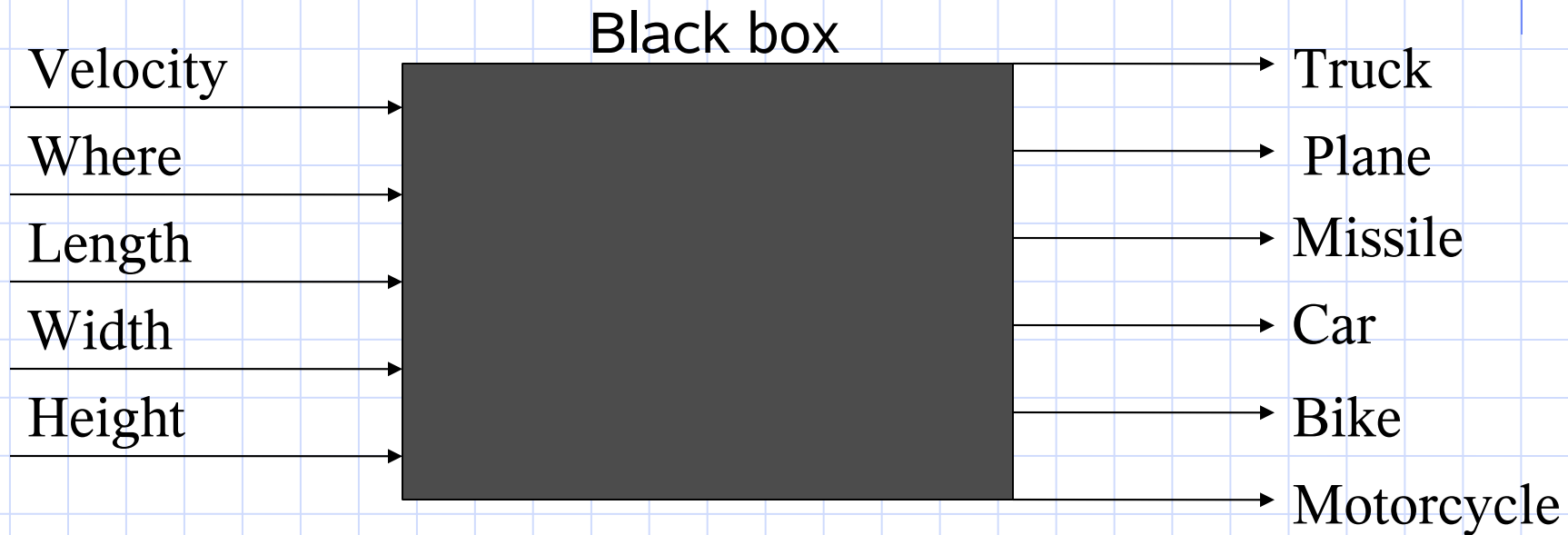


OK, so Descartes beat me to it, but this is in a different realm

Classification Problem

- ◆ Our sensors report:
 - Velocity - fuzzy: low, medium, or high
 - Sky or ground – categorical variable
 - Length – ratio variable
 - Width – ratio variable
 - height – ratio variable

How Do We Classify These?



Creating an Optimal Protein

- ◆ Causal model is not understood
- ◆ Solution: use an artificial neural network (ANN) with a genetic algorithm (GA)
 - Train ANN on known proteins
 - Use trained ANN as fitness function in GA
 - Use GA for exploited search for near optimal protein

Other Applications

◆ Image processing

- Pixel: foreground or background classification
- Non-linear filtering
- Classification
- Pattern recognition

◆ Radar

- Tracker
- Pattern recognition

◆ Medical

- Diagnosis
- Classification
- Pattern recognition

More Applications

◆ Economic

- Credit vetting
- Forecasting
- Fraud detection

◆ Military

- Automatic target recognition
- Steganography
- Image processing

◆ The list goes on

Contents

- Introduction
 - Sample applications
 - Neural network
 - Type of functions
 - Advantages
 - Disadvantages
 - Biological NN
 - How an NN works
 - The neuron
 - Mathematics
1. Compare to regression
 2. Architecture
 3. Training
 4. Dynamic learning & hybrids
 5. Examples
 6. When to use
 7. Future
 8. Information sources

What is a Neural Network?

- ◆ A human Brain
- ◆ A porpoise brain
- ◆ The brain in a living creature
- ◆ A computer program
 - Emulates biological brain
 - Limited connections
- ◆ Specialized computer chip

What is an ANN?

(Artificial Neural Network)

◆ General function approximator

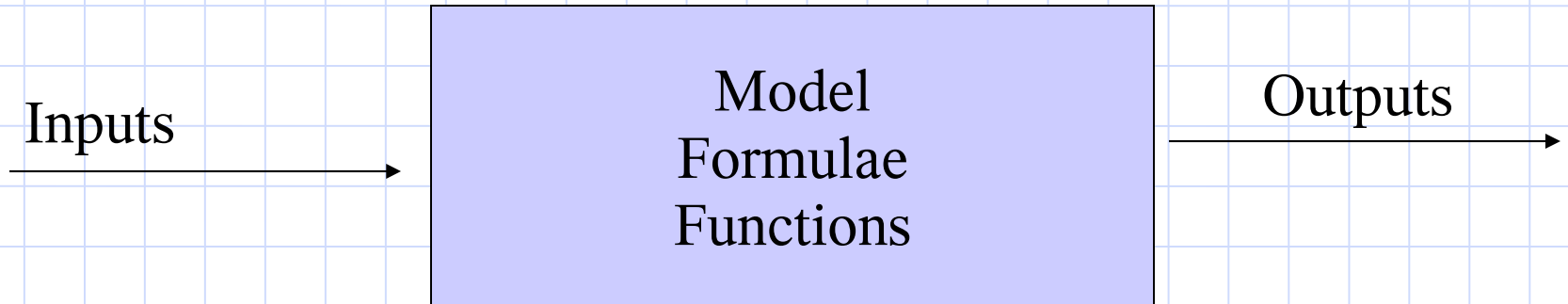
- Imitates performance of original
- Does not duplicate model
- Does provide near or approximate results
- It maps input to output

◆ Data driven

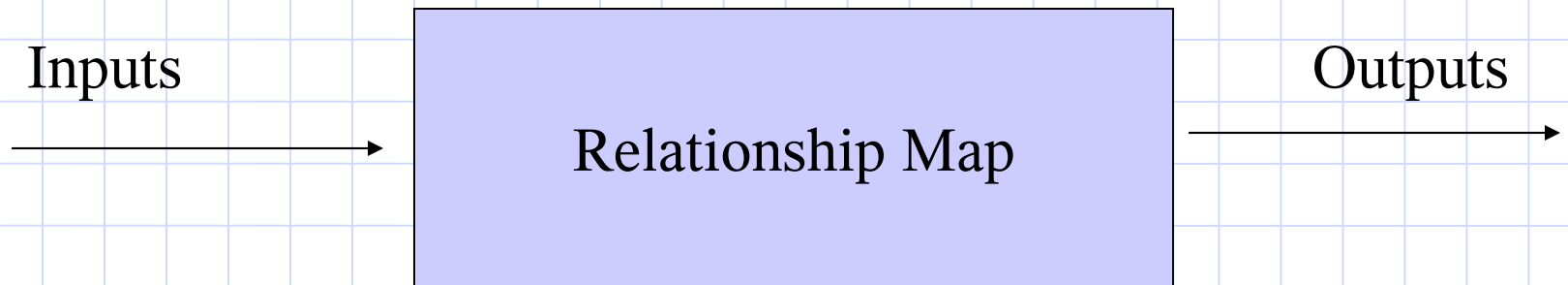
- Does not understand causal model
- Learns input to output relationship
- Learns from supplied training data

Models

Model based



Artificial Neural Network



What Can an ANN Use to Make Connections/Mapping?

- ◆ Learned Information
- ◆ From experience
- ◆ From historical data
- ◆ By example
- ◆ By organization

Four types of Functions

- Prediction and Time Series Forecasting
 - Like regression, but not constrained to linear
- Classification
 - Find which class is the closest match
- Pattern Recognition
 - Fined tuned classification
- Self organizing map for clustering
 - ◆ Not constrained to linear or Gauss Normal distribution
 - ◆ Also used for modeling biological neural network in medical research

Advantages of Neural Network

- ◆ No Expert needed
- ◆ No Knowledge Engineer needed
- ◆ Does not have bias of expert
- ◆ Can interpolate for all cases
- ◆ Learns from facts
- ◆ Can resolve conflicts
- ◆ Variables can be correlated (multicollinearity)

More Advantages

- ◆ Learns relationships
- ◆ Can make good model with noisy or incomplete data
- ◆ Can handle non-linear or discontinuous data
- ◆ Can Handle data of unknown or undefined distribution
- ◆ Data Driven

Disadvantages of Neural Net

◆ Black Box

- don't know why or how
- not sure of what it is looking at

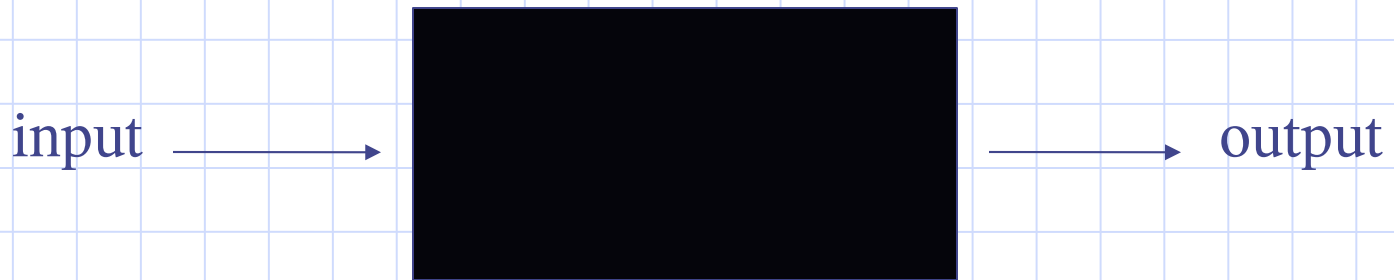
◆ Operator dependent

◆ Don't have knowledge in hand

◆ Just a mapping of input to output, no modeling of system

* *Many of these disadvantages are being overcome*

Black Box



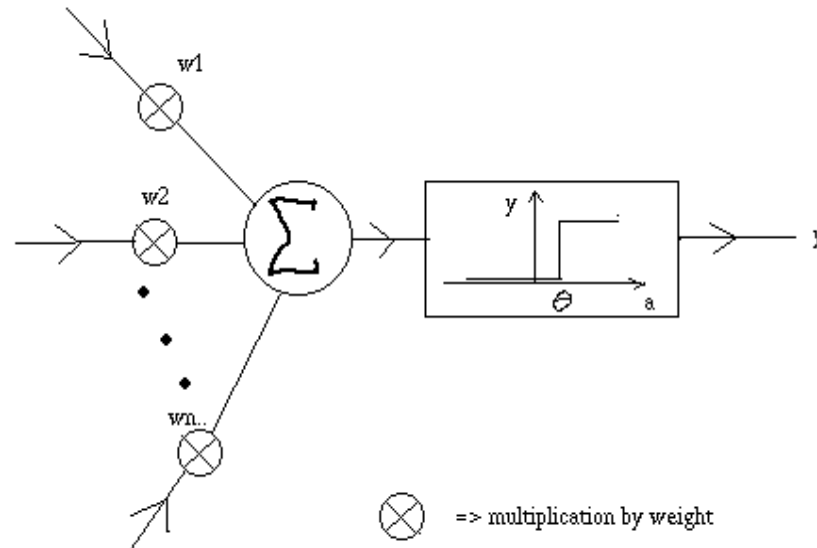
- ◆ What happens inside the box is unknown
- ◆ We can't see into the box
- ◆ We don't know what it knows

Biological Neural Network

- ◆ Human Brain has 4×10^{10} to 10^{11} Neurons
 - ◆ Each can have 10,000 connections*
 - ◆ Human baby makes 1 million connections per second until age 2
 - ◆ Speed of synapse is 1 kHz, much slower than computer (3.0+ GHz)
 - ◆ Massively parallel structure
- * Some estimates are much greater, as much as 100,000

How does a neuron work?

- ◆ It sums the weighted inputs
 - If it is enough, then neuron fires
 - There can be as many as 10,000 or more inputs

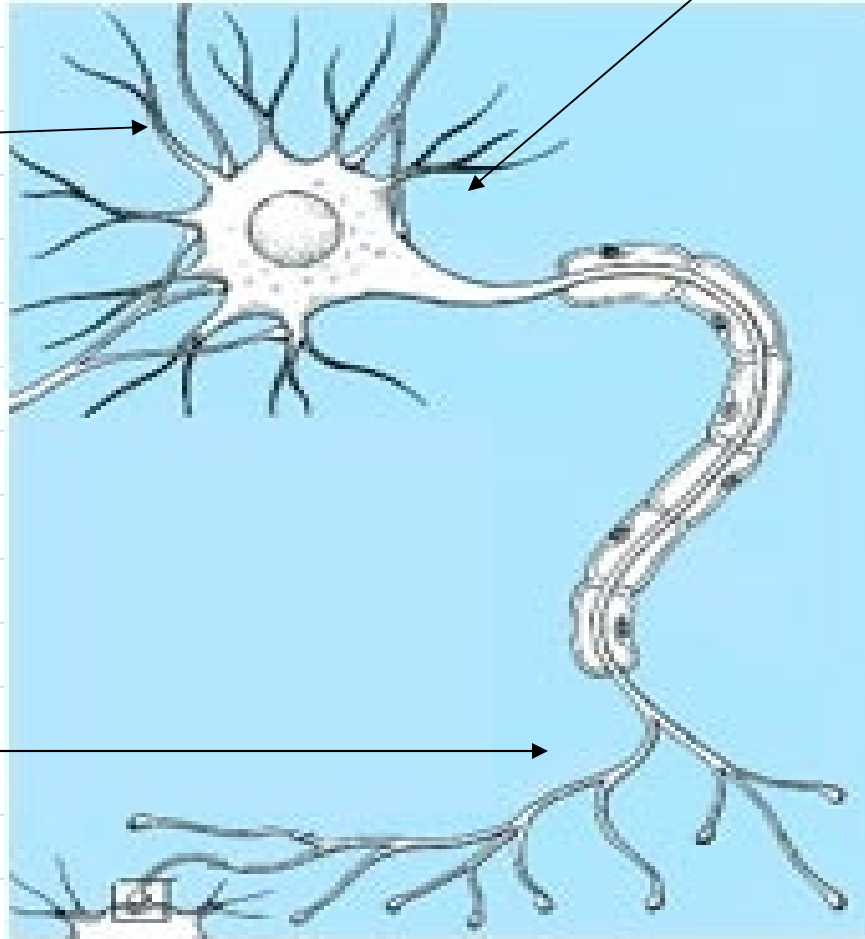


Neuron

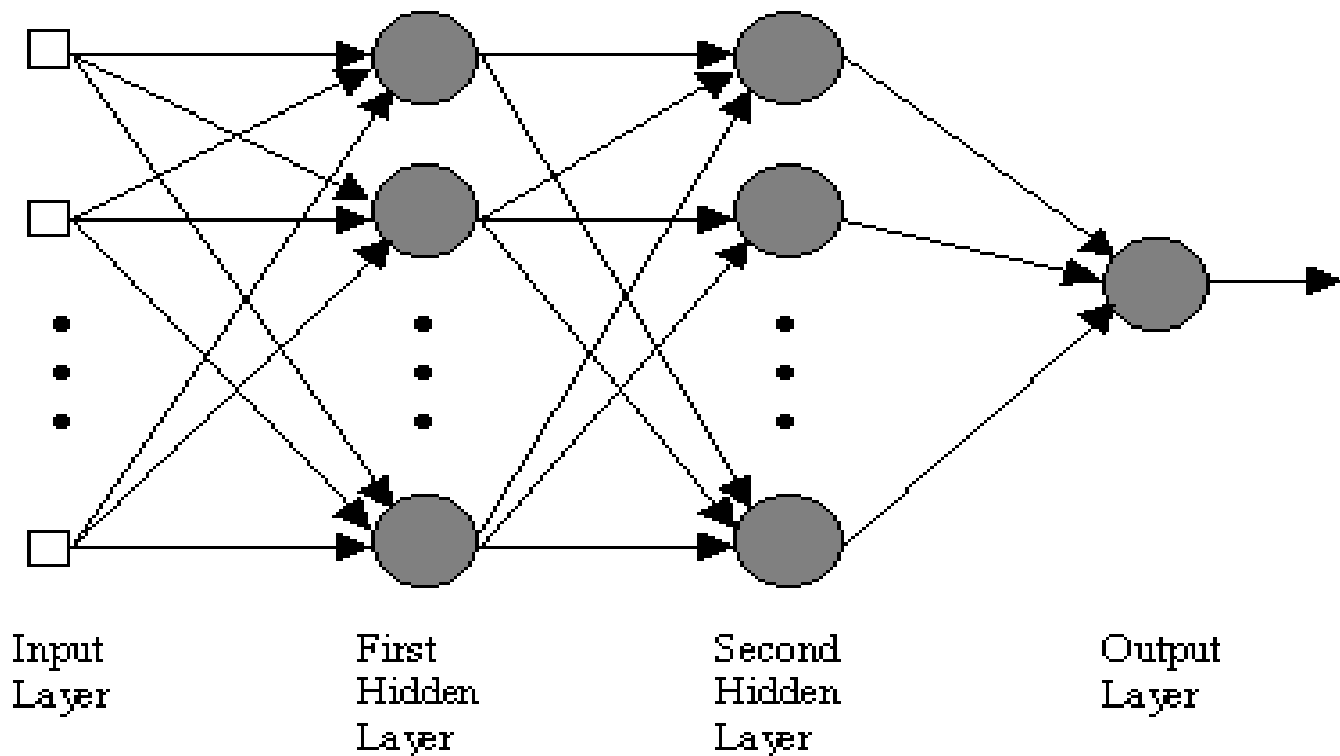
Dendrites
(inputs)

soma (body)

Axons
(outputs)



Neural Network



This is a feed-forward design

Computer Neural Network

- ◆ Von Neumann architecture
- ◆ Serial machine with inherently parallel process
- ◆ Series of mathematical equations
- ◆ Simulates relatively small brain
- ◆ Limited connectivity
- ◆ Closely approximates complex non-linear functions

Neuron Activation

- ◆ Weights can be positive or negative
- ◆ Negative weight inhibits neuron firing
- ◆ $\text{Sum} = W_1N_1 + W_2N_2 + \dots + W_nN_n$
- ◆ If sum is negative, neuron does not fire
- ◆ If sum is positive neuron fires
- ◆ Fire means an output from neuron
- ◆ Non-linear function
- ◆ Some models include a threshold

Neuron Activation

◆ Linear

◆ Sigmoidal

- $1.0/(1.0+e^{-s})$ where $s = \Sigma$ inputs
- 0 or +1 result

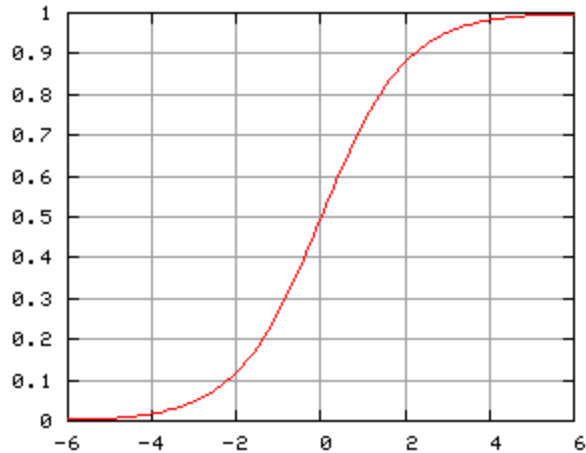
◆ Hyperbolic Tangent

- $(e^s - e^{-s}) / (e^s + e^{-s})$ where $s = \Sigma$ inputs
- -1 or +1 result

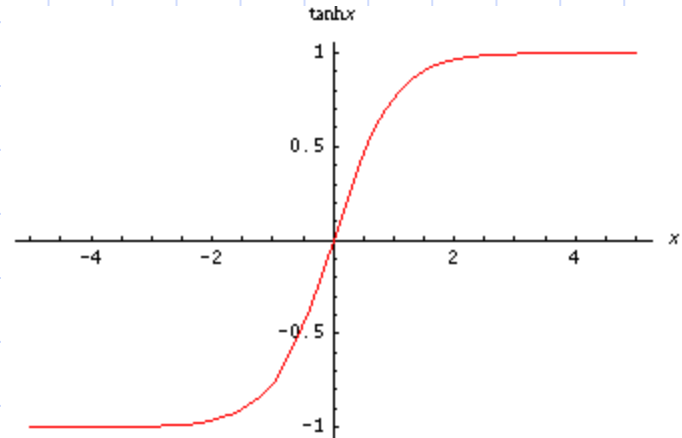
◆ Also called squashing or clamping function

- Because it takes a large value and compresses it
- Adds the non-linearity to the process

Activation Functions



Sigmoidal Function
Goes from 0 to 1
Hard to be at extreme



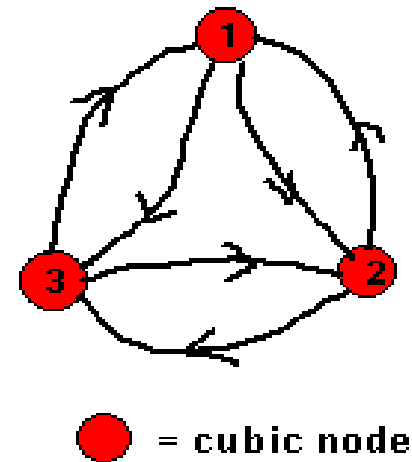
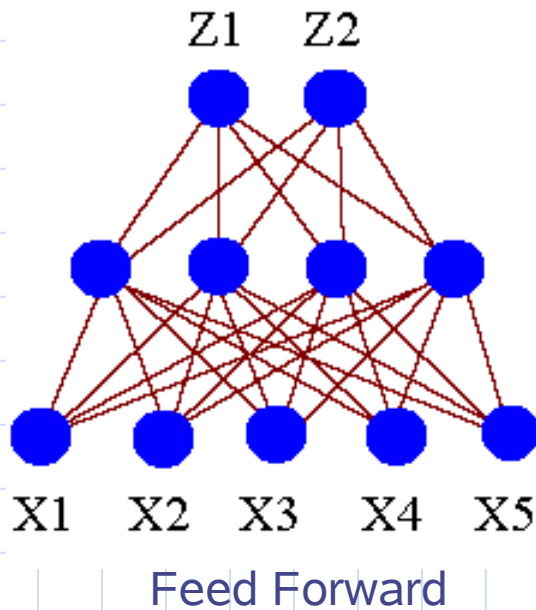
Hyperbolic Tangent
Goes from -1 to 1
Hard to be at extreme

Neuron Math

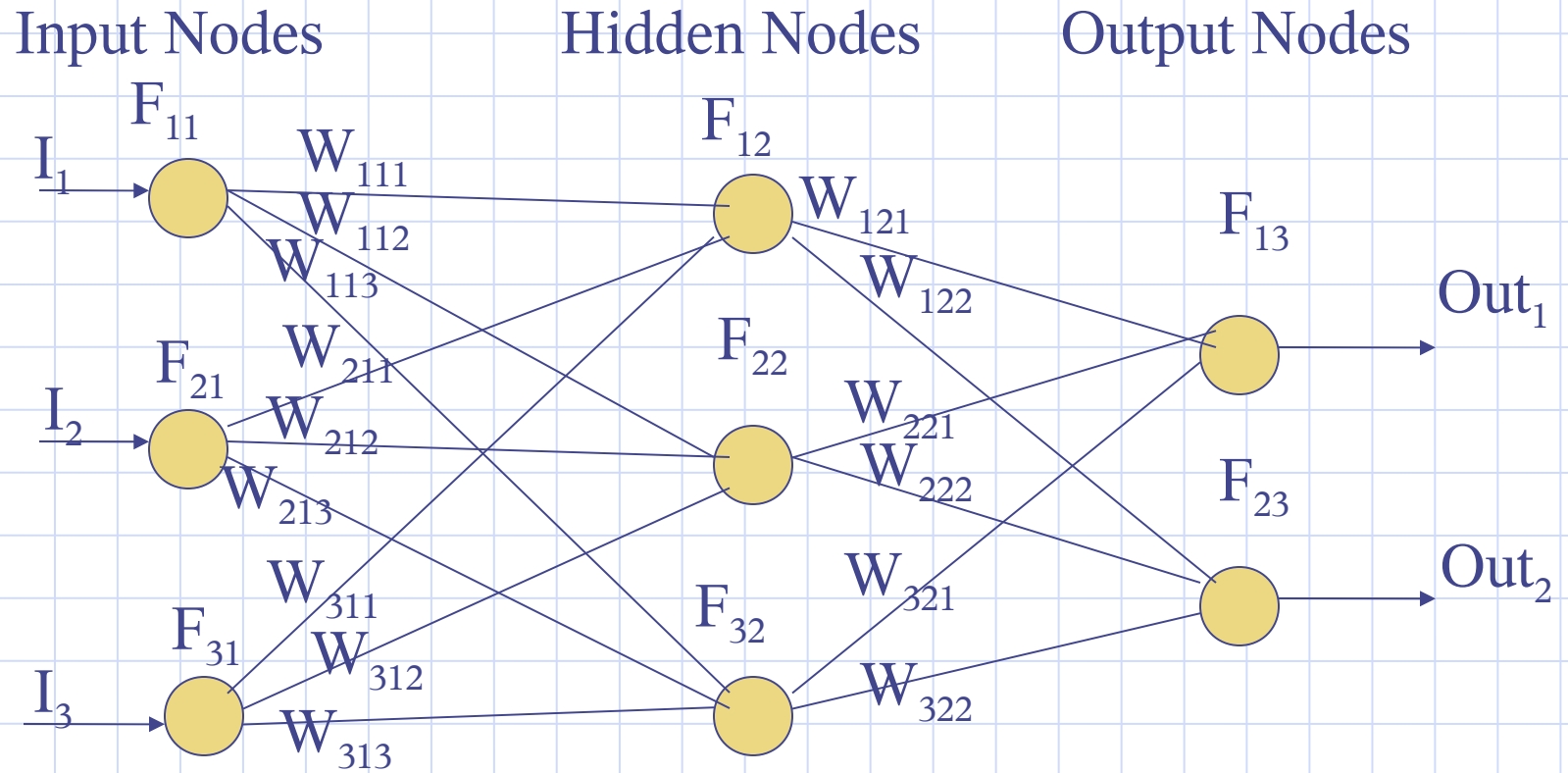
- ◆ Don't try for 0 or 1
 - Use 0.1 and 0.9 instead for logistic
 - Use -0.9 and $+0.9$ for hyperbolic tangent
- ◆ Real plane math
- ◆ Complex domain math
 - Quite often outperforms systems using real domain math
 - Better for signal & image processing

What does the network look like?

- ◆ This is a computer model, not biological
- ◆ Left has 11 neurons, sea slug has 100



Small Neural Network



Regression?

- ◆ With linear activation, this is but parallel regression
- ◆ With sigmoid or H-Tan, this is a parallel logistic regression
- ◆ An ANN with zero hidden nodes, one output, and linear activation *is* OLS regression if the objective function is minimizing SSE (sum of squared error)

Mathematical Equations

◆ Input to Hidden₁₂ = H_1

$$\text{◆ } H_1 = [(I_1 * F_{11}) * W_{111}] + [(I_2 * F_{21}) * W_{211}] + [(I_3 * F_{31}) * W_{311}]$$

$$\text{◆ } H_2 = \dots\dots$$

$$\text{◆ } H_3 = \dots\dots$$

$$\text{◆ } \text{Out}_1 = [(H_1 * F_{12}) * W_{121}] + [(H_2 * F_{22}) * W_{221}] + [(H_3 * F_{32}) * W_{321}]$$

Matrix Math

◆ Makes it very simple!

$$F(A \times W) = \text{Out}$$

In Fortran:

```
out = Active(matmul(input, weights))
```

Where F or Active is the activation function

Can also use Matlab/Mathematica but it will compute more slowly as they are interpretive

Comparison to Regression

- ◆ OLS with 3 independent and 1 dependent variables would have a maximum of 3 coefficients and 1 intercept
- ◆ With 2 dependent variables, it would require Canonical Correlation (general linear model) and the same number of coefficients
- ◆ ANN (with one hidden layer) has 15 coefficients (weights) and activation functions can be non-linear
- ◆ Multicollinearity is not a problem in an ANN

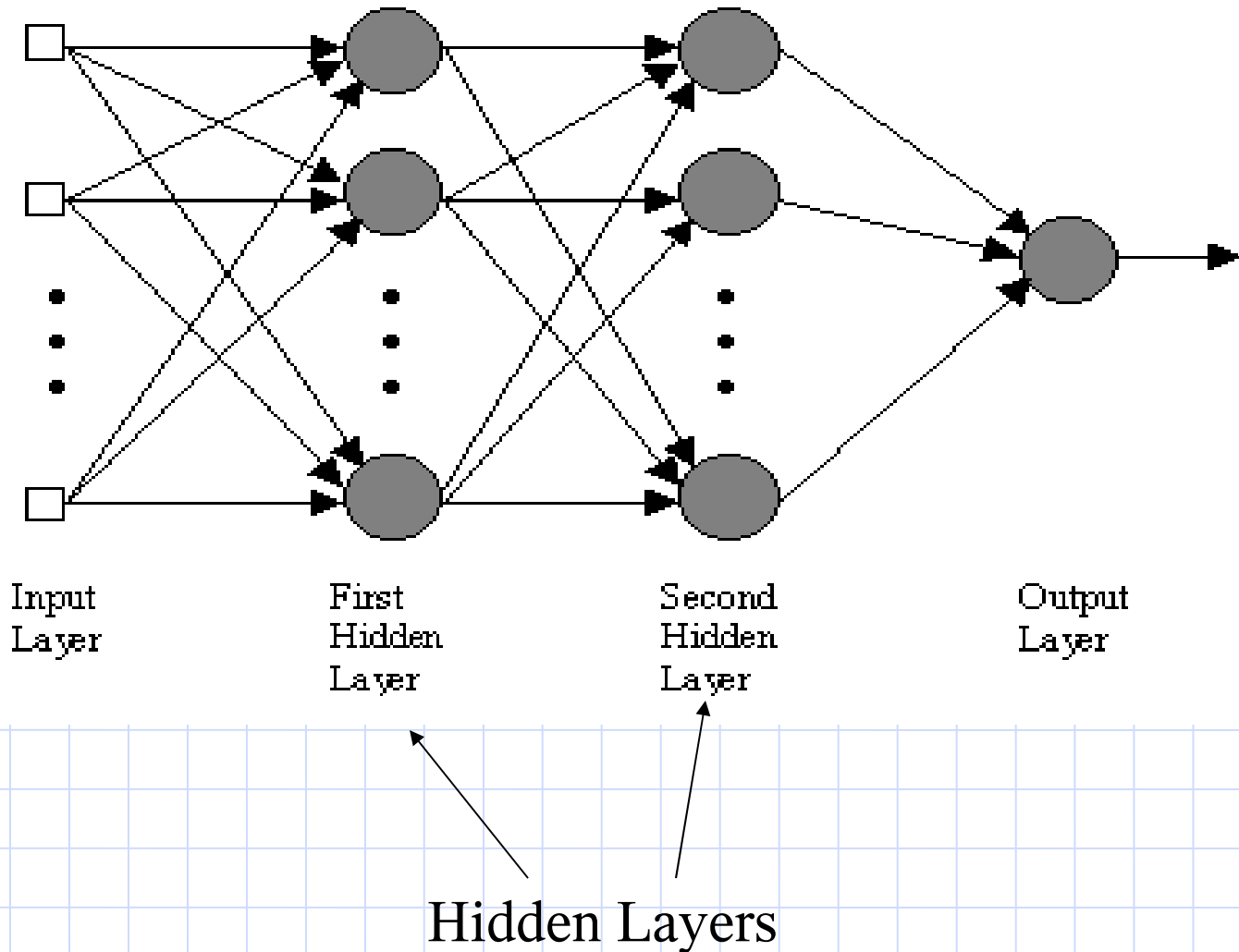
Inputs

- ◆ One per input node
- ◆ Ratio
- ◆ Logical
- ◆ Dummy
- ◆ Categorical
- ◆ Ordinal
- ◆ Fuzzy (PNL)
- ◆ Functional Link Network
 - Interaction variable
 - Transformed variable

Hidden Layer(s)

- ◆ Increase complexity
- ◆ Can increase accuracy
- ◆ Can reduce degrees of freedom
 - Need larger data set
- ◆ Presently architecture up to programmer
- ◆ Source for error
- ◆ In future will be more automatic
 - Some literature describes this

Hidden Layer(s)

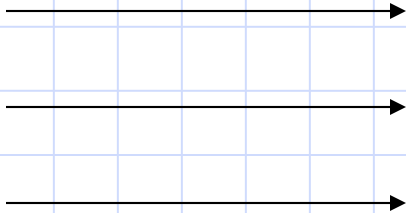


Outputs

- ◆ One for single dependent variable
- ◆ Multiple
 - Prediction
 - Classification
 - Pattern recognition

Outputs

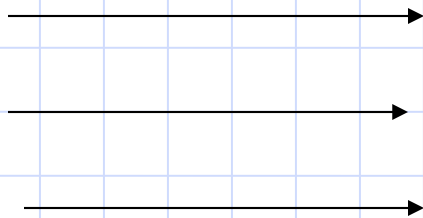
Inputs



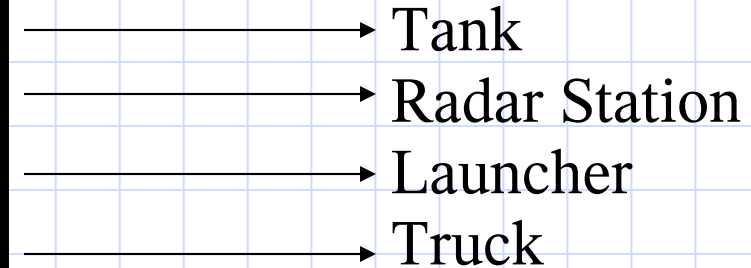
Single output



Inputs



Multiple Outputs



Macro View of Training

- ◆ Setting all of the weights
- ◆ To create optimal performance
- ◆ Optimal adherence to training data
- ◆ Really an optimization problem
 - Optimal methods depends on many variables
 - See optimization lecture
- ◆ Need objective function
- ◆ Beware of local minima!

Supervised or Not

◆ Supervised

- Train it with examples
- And give it the answers
- Much like school

◆ Unsupervised

- Give it examples
- Do NOT give it answers
- It organizes the data by similarities

Training

◆ Supervised

Pattern 1 → Answer 1

Pattern 2 → Answer 2

Pattern 3 → Answer 3

◆ Unsupervised

Pattern 1

Pattern 2

Pattern 3

Optimization Methods to Set the Weights

- ◆ Back Propagation (most popular)
- ◆ Gradient Descent
- ◆ Generalized reduced gradient (GRG)
- ◆ Simulated Annealing
- ◆ Genetic Algorithm
- ◆ Two or more output nodes
 - Multi objective optimization (hard problem)
- ◆ Many more methods

Training Data Set

- ◆ Need more observations than weights
 - Positive number degrees freedom
- ◆ More observations is usually better
 - Lower variance
 - More knowledge
- ◆ Watch aging of data
- ◆ Data must be representative of population

Data Window

◆ Rolling Window

- Rolls forward including all data behind
- Constant starting point with ever increasing size

◆ Moving Window

- Deletes the oldest as it adds the newest
- Constant size with ever increasing starting point
- Necessary when underlying factors change

Rolling vs. Moving Window

◆ Rolling Window



◆ Moving Window

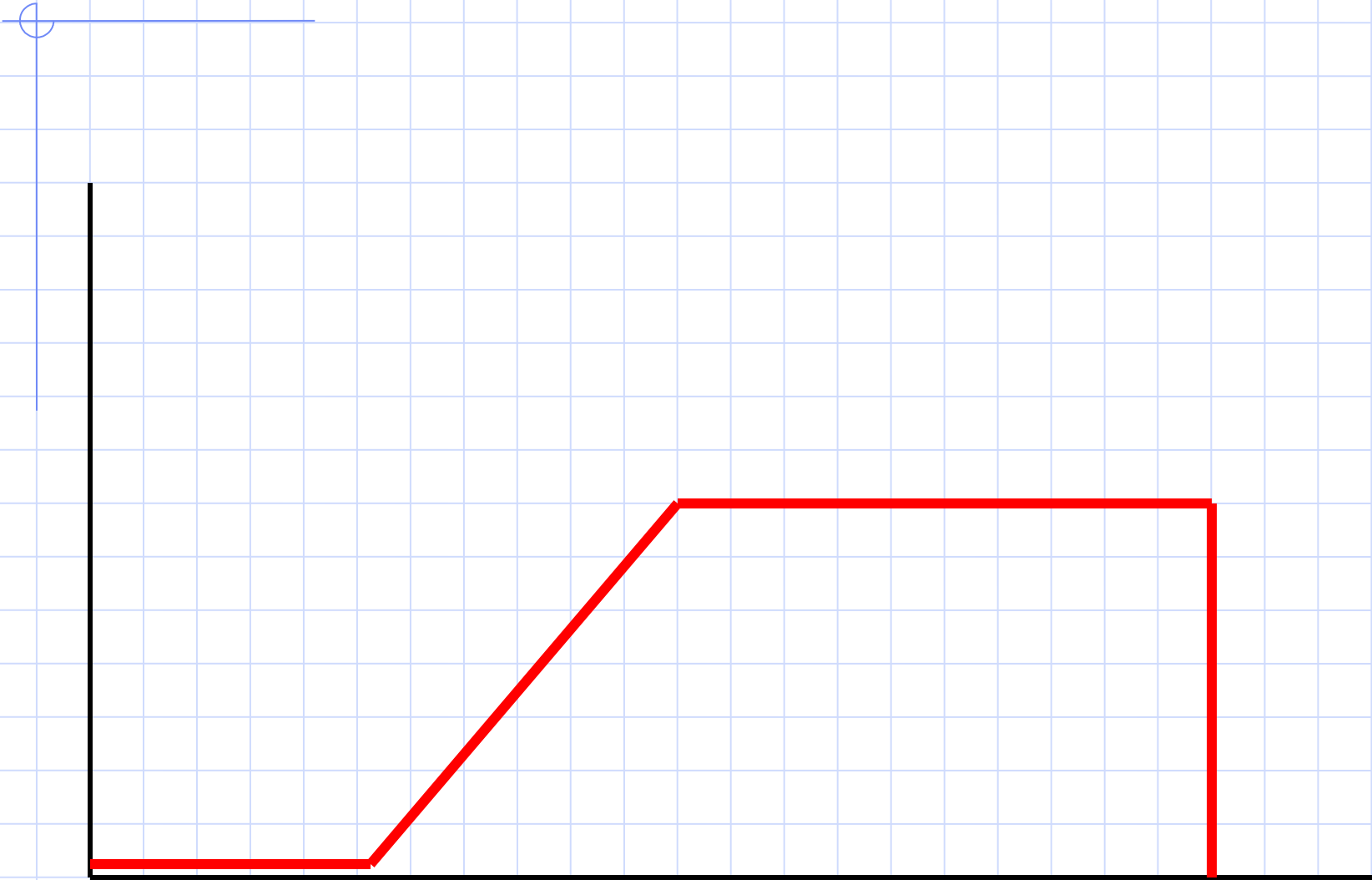


Data Window Continued

◆ Weighted Window

- Morantz, Whalen, & Zhang
- Superset of rolling & moving window
- Oldest data is reduced in importance
- Has reduced residual by as much as 50%
- Multi factor ANOVA shows results significant in majority of applications with real world data

Weighted Window



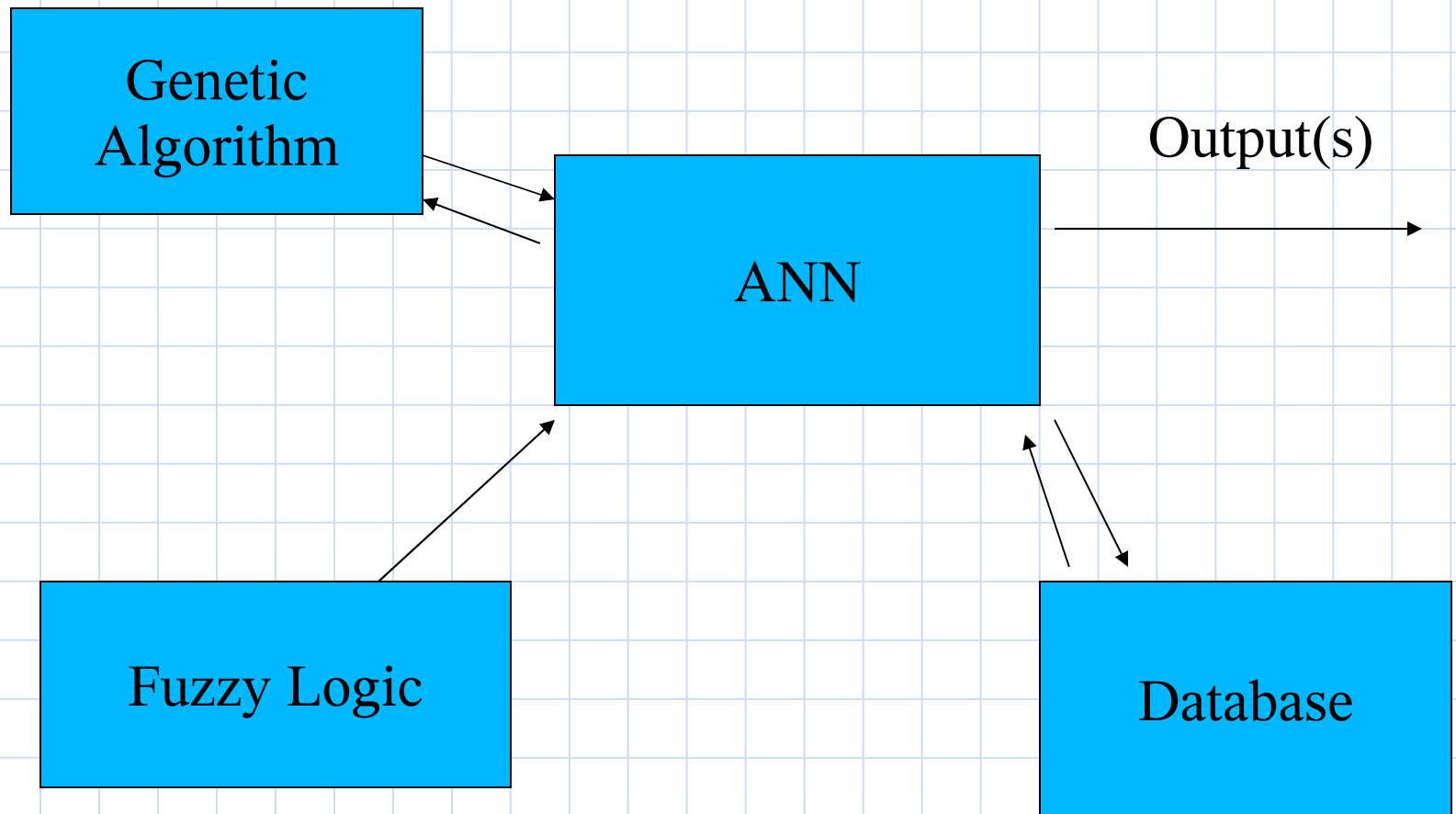
Dynamic Learning

- ◆ Also called reinforcement learning
- ◆ Continuous learning
 - From mistakes and successes
 - From new information
- ◆ Shooting baskets example
 - Too low. Learned: throw harder
 - Too high. Learned: throw softer, but not as soft as before
 - Basket! Learned: correct amount of “push”
- ◆ Loaning \$10 example

Hybrids

- ◆ Combine several systems
 - GA and ANN
 - ANN with fuzzy, GA, & database
 - Many possibilities
- ◆ Uses more methods than just one type
- ◆ Can seed system with expert knowledge and then update with data
- ◆ Sometimes hard to get all parts to work together
- ◆ Harder to validate model

Hybrids



Example

- ◆ You go some place that you have never been before, and get “bad vibes”
 - Atmosphere, temperature, lighting, smell, coloring, numerous things
- ◆ For some reason, brain associates these together, possibly some past experience
- ◆ Gives you “bad feeling”

Additional Examples

- ◆ Military: submarine, tank, & sniper detection
- ◆ Security
- ◆ Classify stars & planets
- ◆ Data mining
- ◆ Natural language recognition
- ◆ OCR including Kanji

My Favorite Examples

- ◆ Fire control for ABL (air borne Laser)
- ◆ ANN with GA hybrid
- ◆ With real constraints
- ◆ Initially trained from panel of experts
- ◆ Ran in simulation
 - Learned from mistakes
 - Retrained after each set of sorties
 - Improved performance (less leakers)
 - From Stroud, IEEE Transactions on Neural Networks

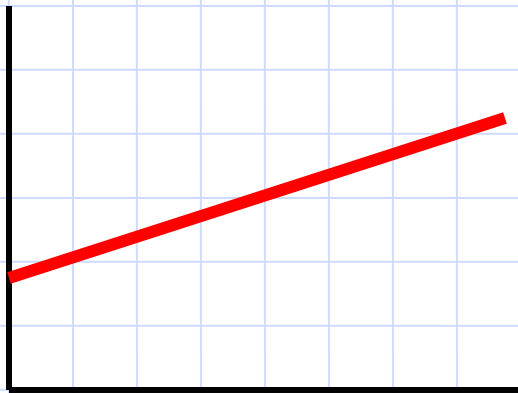
The Other Favorite Example

- ◆ The brain of a bat
 - Size of a plum
 - Controls voluntary & involuntary processes
 - Controls sonar system and navigation
 - ◆ Outperforms our best navigation systems
 - ◆ Bat can fly through moving electric fan

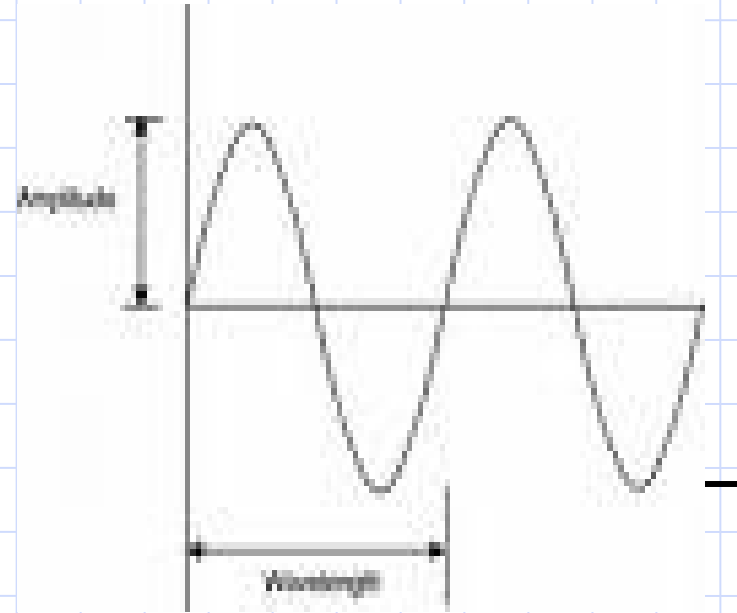
When to Use?

- ◆ Look at the data
- ◆ Is data linear over range of interest?
- ◆ Is Regression accurate enough?
 - Occam's razor says to use it if it is
- ◆ Is data non-linear and/or discontinuous?

What to Use



Regression is fine



Use the ANN here
Regression won't
Fit it well

ANN Chip

- ◆ Original funding was from TEAMA
 - Goal was for use as intelligent appliance
 - ◆ Toaster learned how you like your toast
 - ◆ Coffee pot learned how you want coffee
- ◆ JPL
 - Stack chip
 - For vision applications

Future

- ◆ Rule extraction
- ◆ Hybrids
- ◆ Dynamic learning
- ◆ Parallel processing (it is here)
- ◆ Dedicated chips (ZISC chip)
- ◆ Bigger & more automatic
- ◆ Machine Cognition

About Me

- ◆ I am a Decision Scientist
- ◆ I work on methods to make intelligent *High Quality* decisions
- ◆ Neural networks are a tool in my tool-box
- ◆ I use them like regression, except that they can be non-linear
- ◆ Not the case of only having a hammer and all problems looking like a nail.

Information Sources

- ◆ www.machine-cognition.com
- ◆ IEEE Transactions on Neural Networks
- ◆ IEEE Intelligent Systems Journal
- ◆ IEEE Computational Intelligence Society
- ◆ AAAI American Association for Artificial Intelligence
- ◆ www.ieee.org
- ◆ Internet

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