

2013 Convention

new solutions for a new world

31 Oct - 1 Nov 2013

Sandton, Johannesburg

ACTUARIAL
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An Introduction to Artificial Neural Networks

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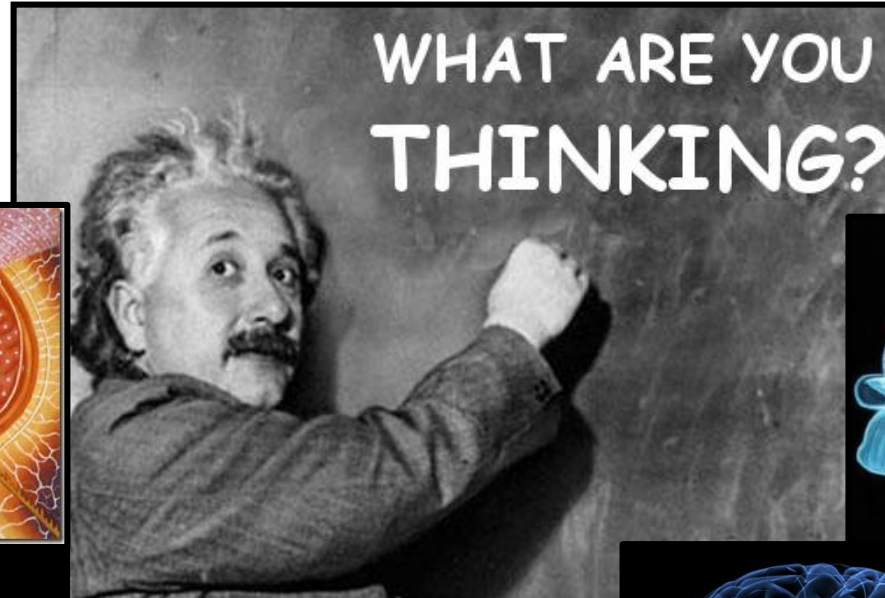
2013 Convention

31 Oct & 1 Nov

Two components to my research



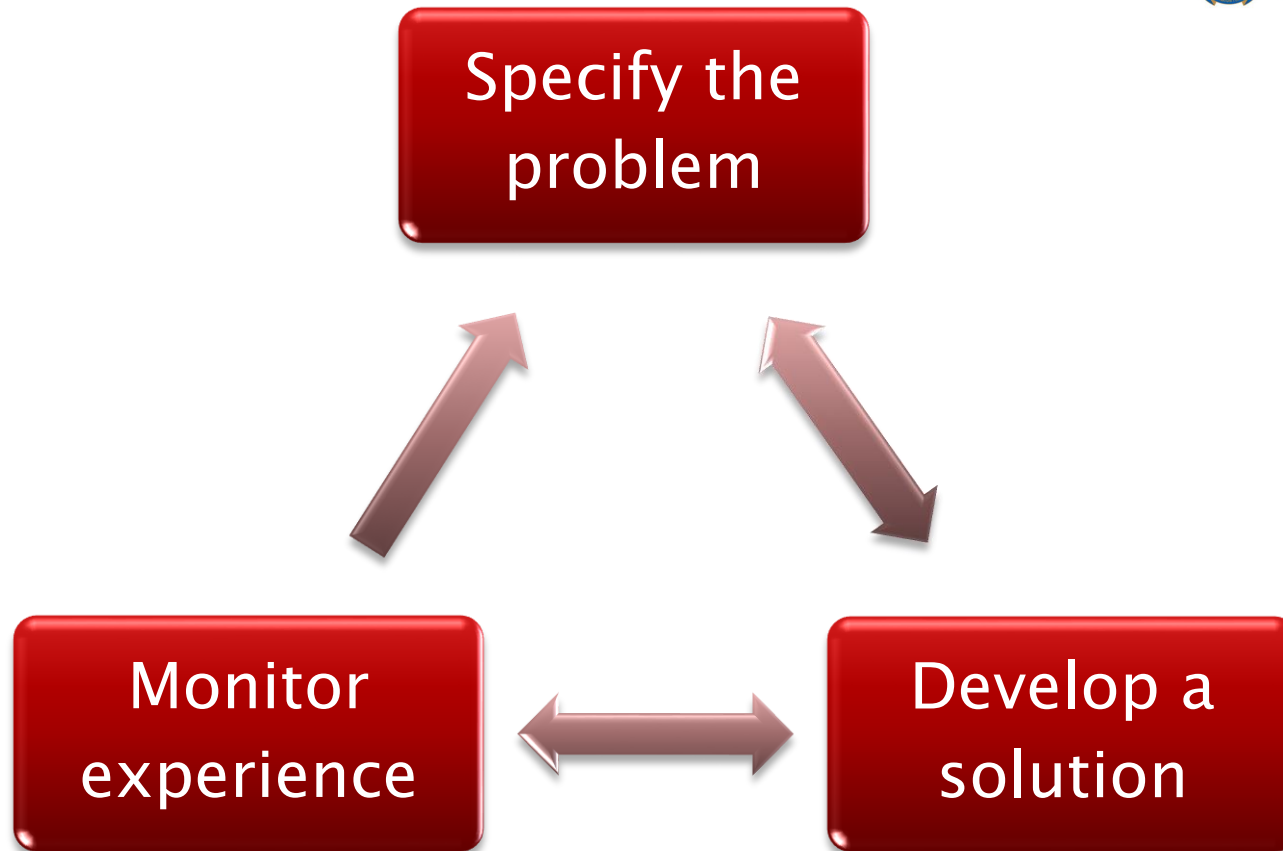
Intrigue into the human mind



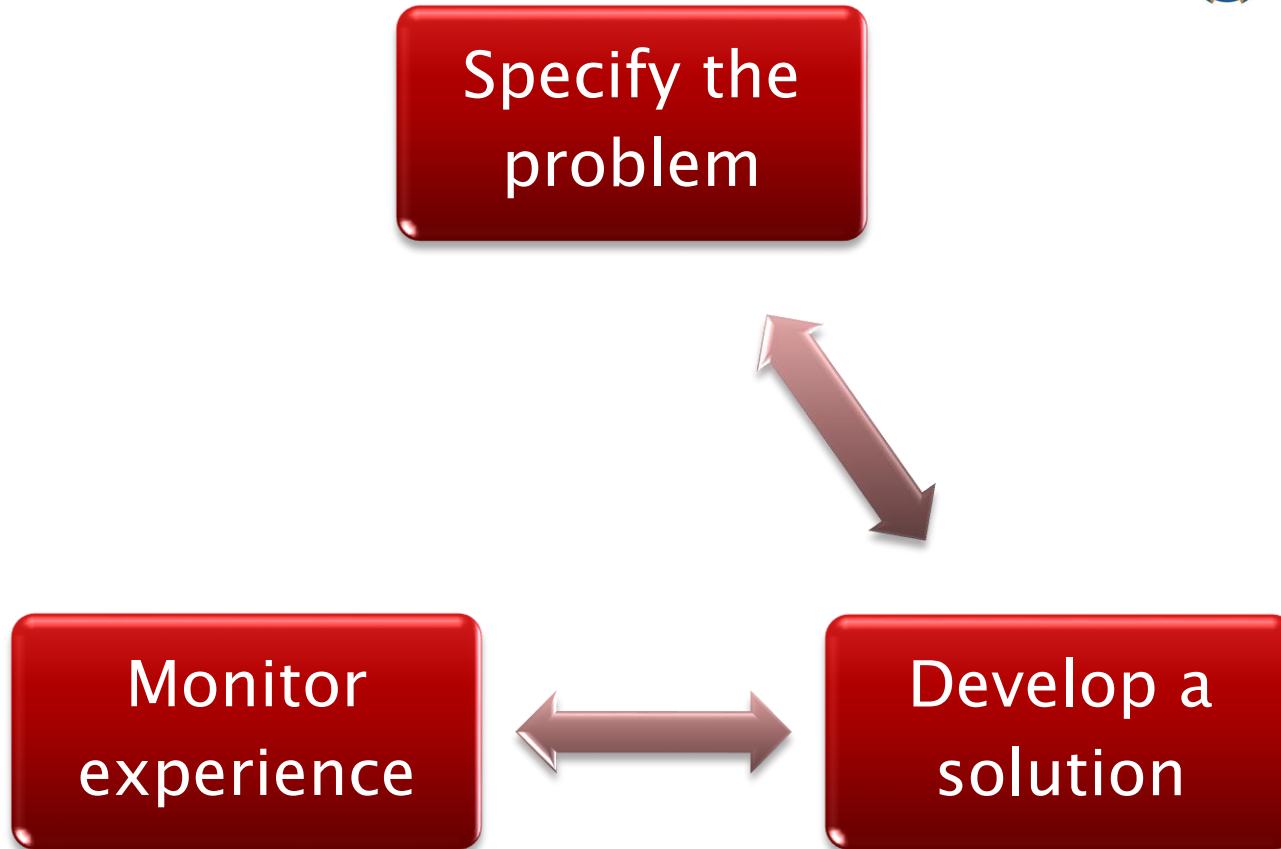
Two components to my research



Actuarial Control Cycle



Actuarial Control Cycle



Road Map

1. What are ANNs?
2. Where do they come from?
3. What do they look like?
4. How do they learn?
5. Problems they can solve?
6. Where have they been applied?
7. Summary



1. What are Artificial Neural Networks?



- ▶ Dynamic modelling technique
- ▶ Replicates the human brain's logic
 - Invented to further understanding of the human brain
- ▶ Has the ability to detect and project non-linear trends in data
- ▶ Can adapt and learn from changing environments

2. Where do Artificial Neural Networks come from?



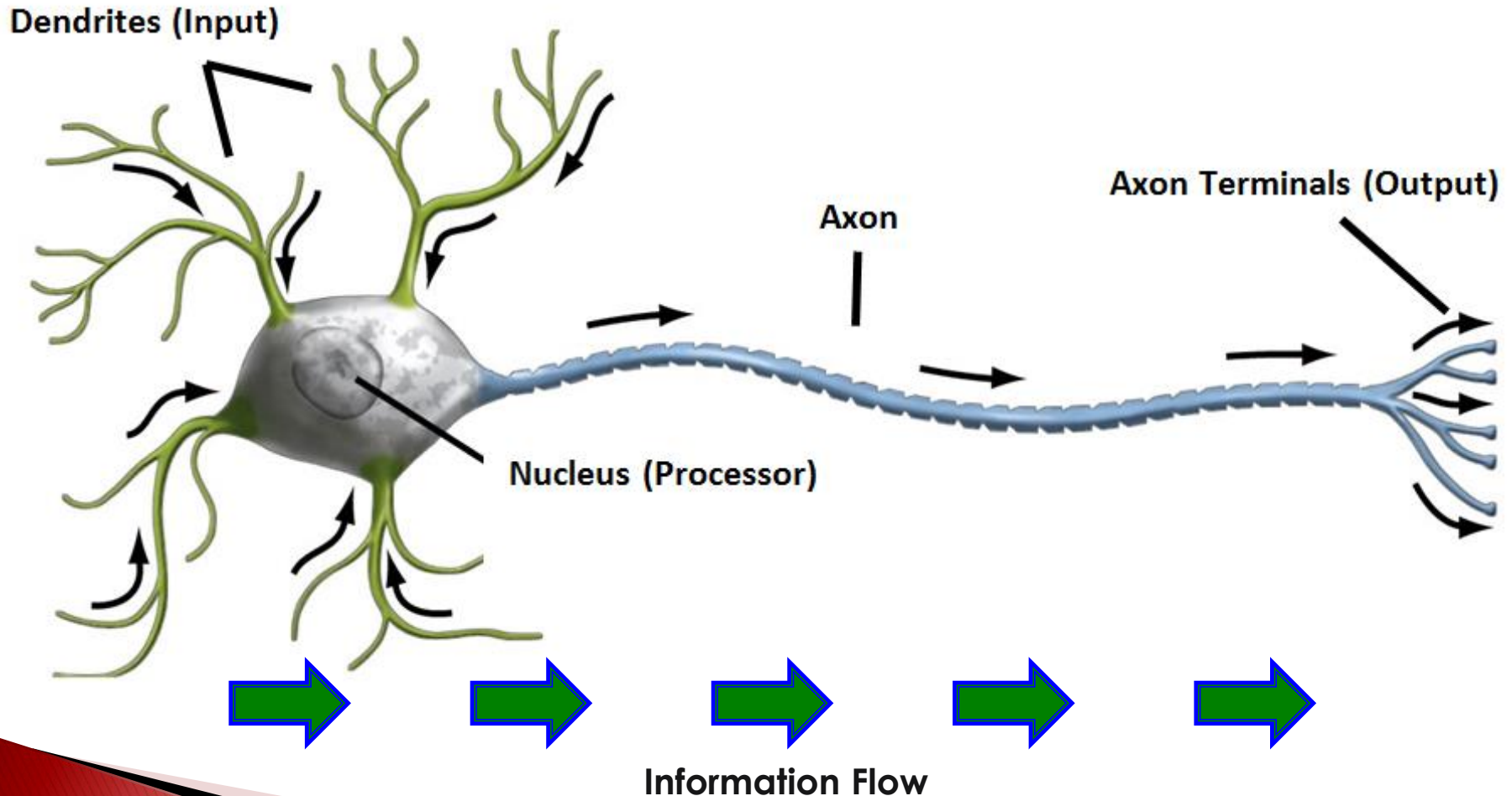
The Human Brain

- ▶ 10 billion neurons
 - Neurons = processors
- ▶ 60 trillion interconnections
 - Massive complex system
- ▶ Characteristics
 - Massive parallel system
 - Fault tolerance
 - Adaptability

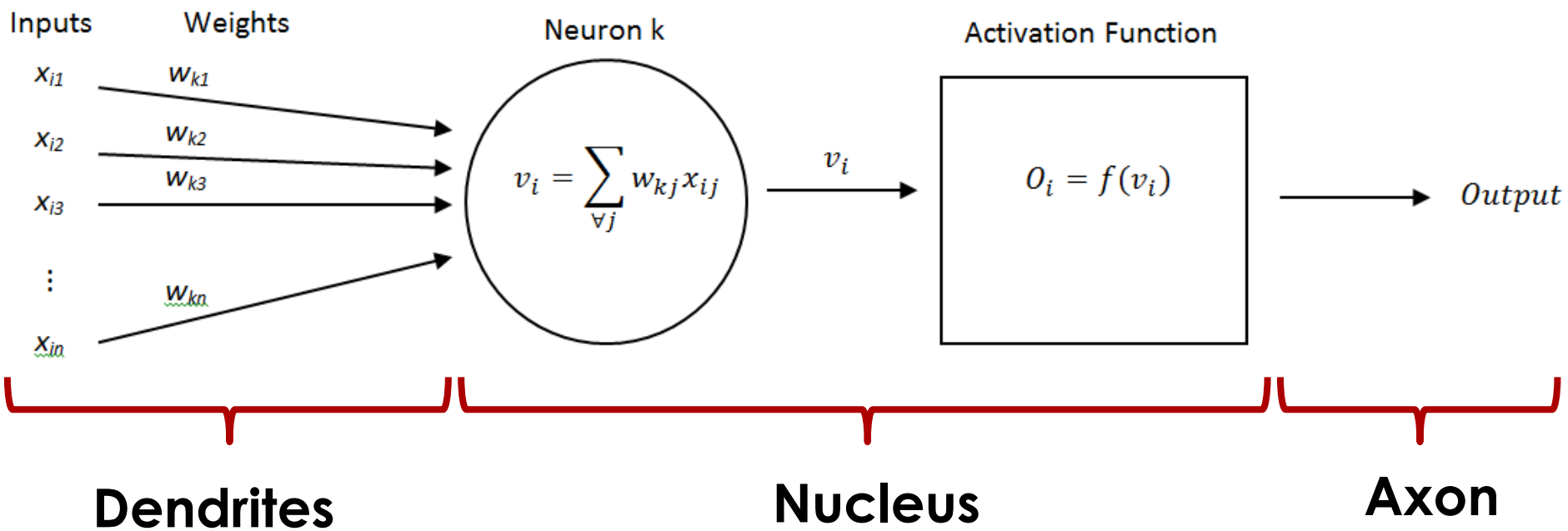


A Biological Neuron

Diagram of a Neuron

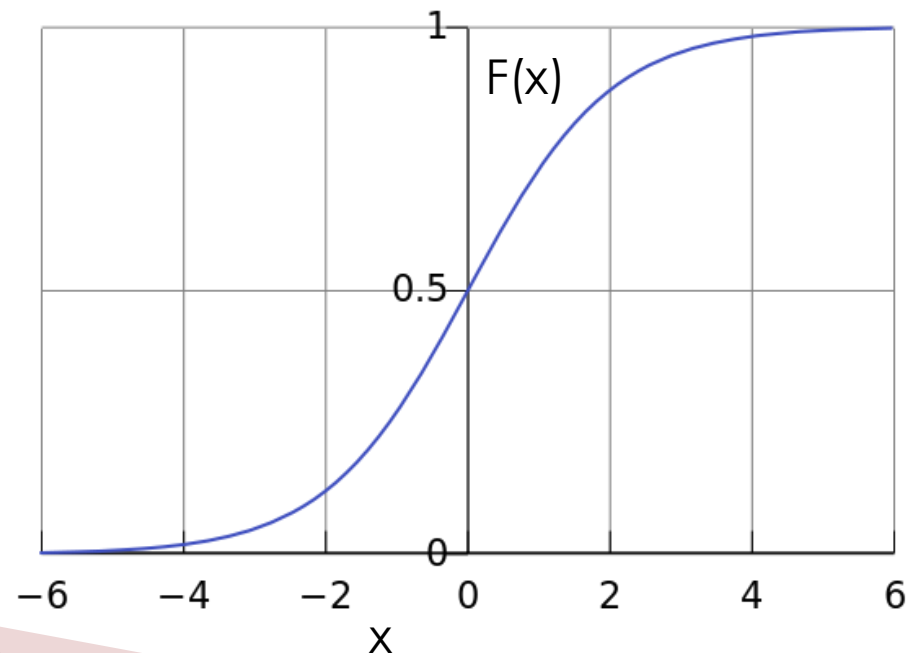


An Artificial Neuron

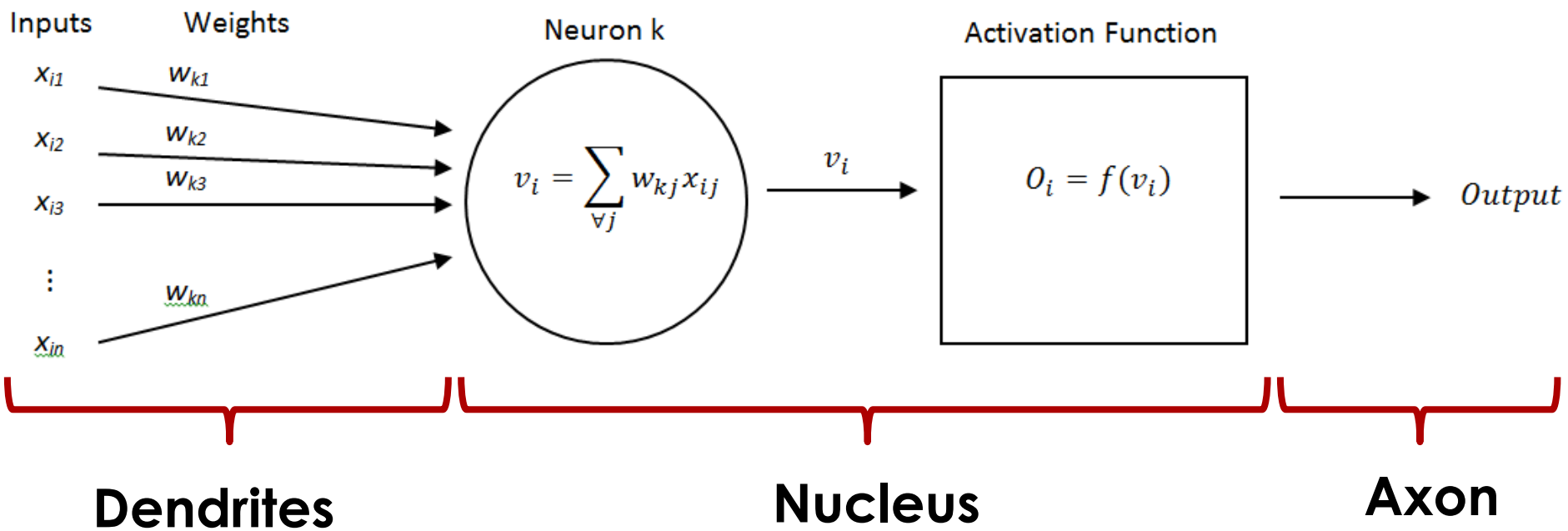


An Activation Function

- ▶ Type of soft switch
- ▶ Sigmoid function $f(x) = \frac{1}{1 + e^{-x}}$
 - High values of x has values close to 1
 - Low values of x has values close to 0



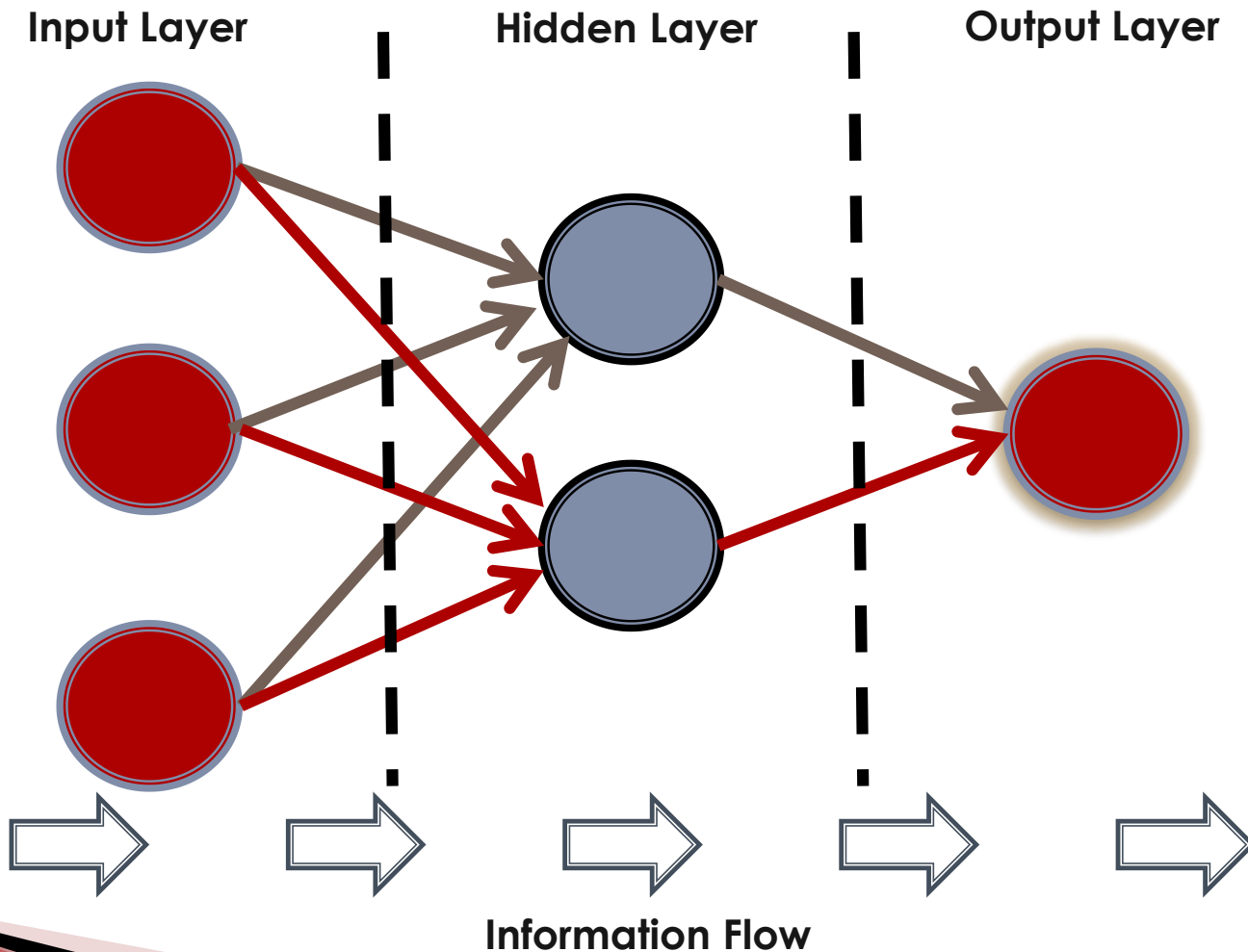
An Artificial Neuron



3. What do they look like?

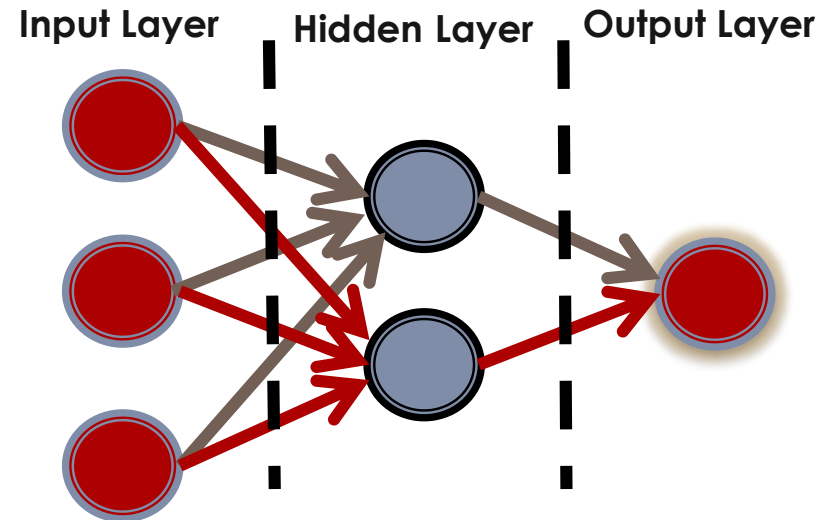
- ▶ Several types of ANNs:
 - Recurrent
 - *Feed-Forward*
 - Bridged
 - Fully Connected Cascade
 - *Multi-Layered Perceptron*
- ▶ We will consider the ***Feed-Forward Multi-Layered Perceptron***
 - Simplest
 - Increase in complexity does not warrant increased accuracy in most practical cases

Feed-Forward Multi-Layered Perceptron (FF MLP)



Feed-Forward Multi-Layered Perceptron (FF MLP)

- ▶ Input layer
 - Data enters the system
 - Each neuron represents a variable
- ▶ Hidden layers
 - Can be several layers
 - Several neurons in each layer
 - A single layer is sufficient
- ▶ Output layer
 - Results of system
 - Can have several results

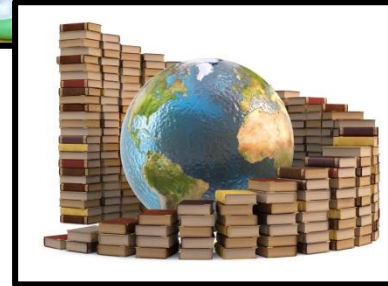
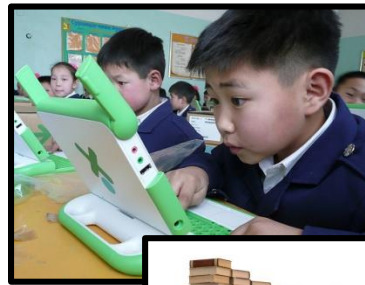


4. How do they learn?



Human Learning

People learn in 5 different ways:



- ▶ Error-Correction

- ▶ Memory

- ▶ Competitive

Supervised Learning

- ▶ Hebbian

- ▶ Boltzmann

Unsupervised Learning

Error-Correction Learning



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How?

We have gained knowledge from:

- Teachers
- Parents
- Guardians
- Past experience



Made a decision that minimised error from past experience



Mathematically How?

- Choose an activation function

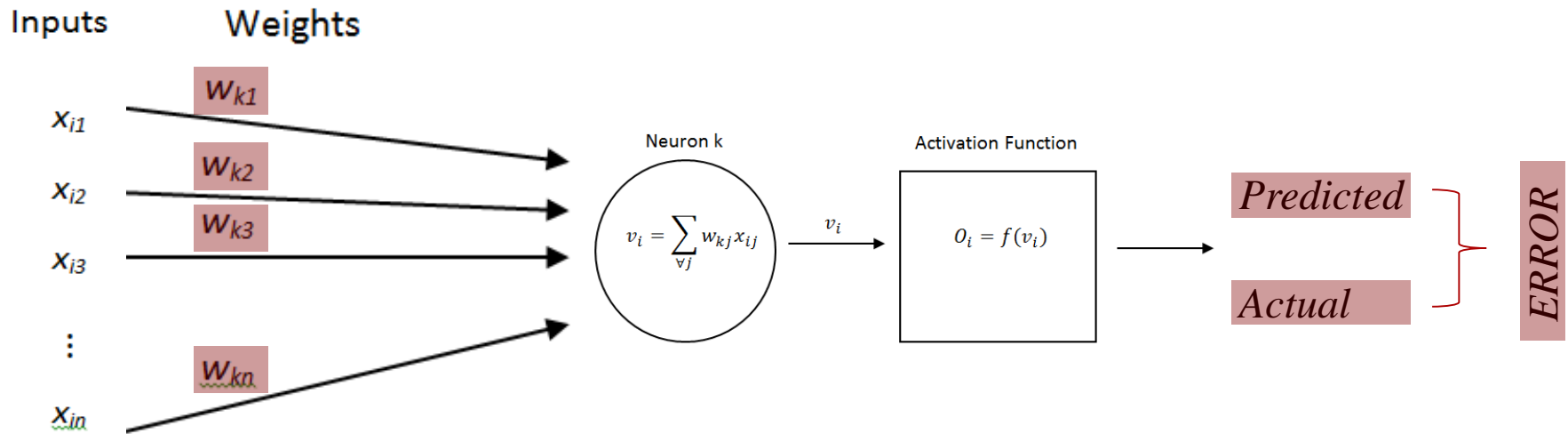
$$A()$$

- Choose an error function

$$E[(A() - actual)^2]$$

- Calculate total error using all past data
- Change the weights using a training algorithm
 - Based on gradient descent
- Specify a number of training cycles until error is small enough

Learning Mechanism



Applying Learning

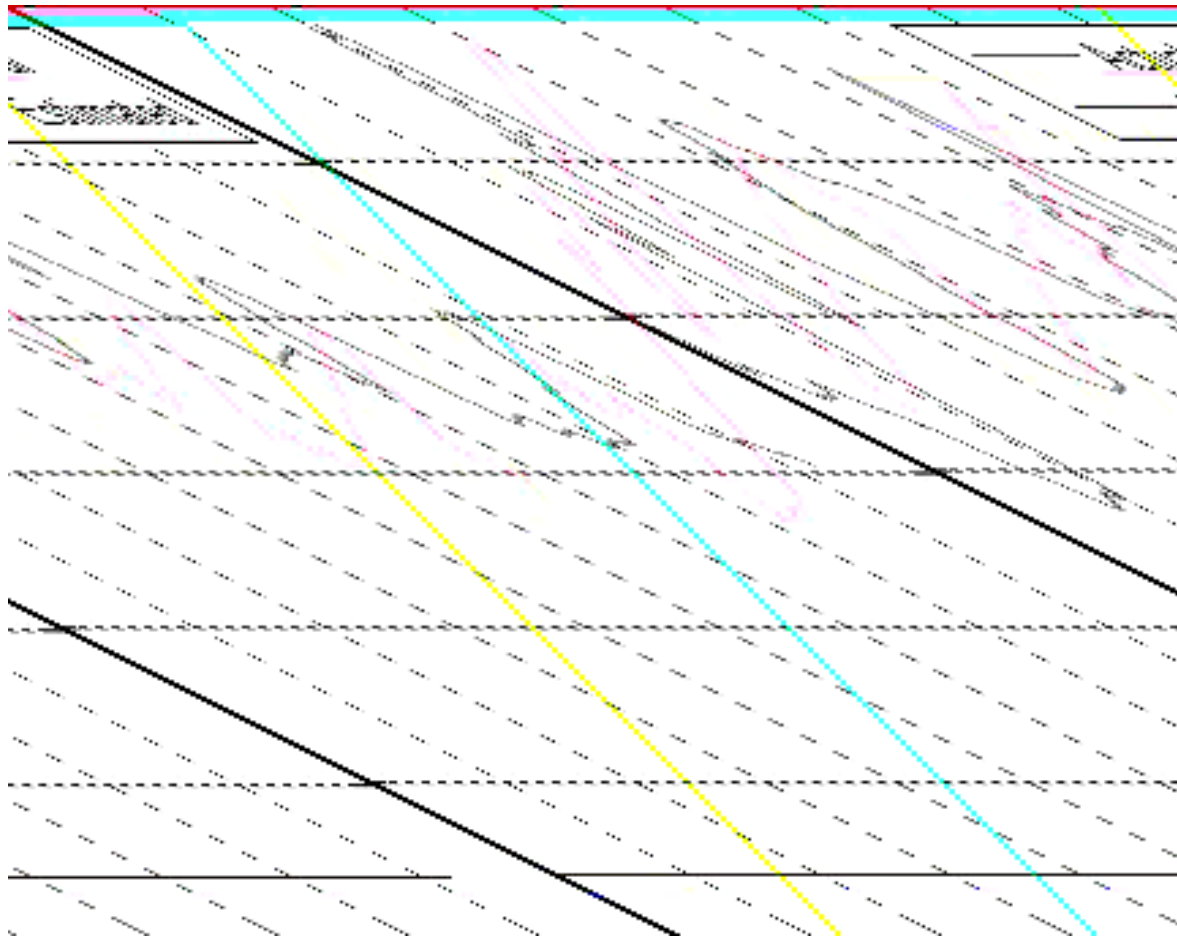
- ▶ Average return on Money Market 1975 - 2010
- ▶ Sources of data:
 - Firer, C and Staunton, M., 2002. 102 Years of South African financial market history. Investment Analysts Journal, 56, 57-65
 - Firer, C. and McLeod, H., 1999. Equities, Bonds, Cash and Inflation: Historical performance in South Africa, 1925 - 1998. Investment Analysts Journal, 50, 7-28.

Learning

Actual vs Expected



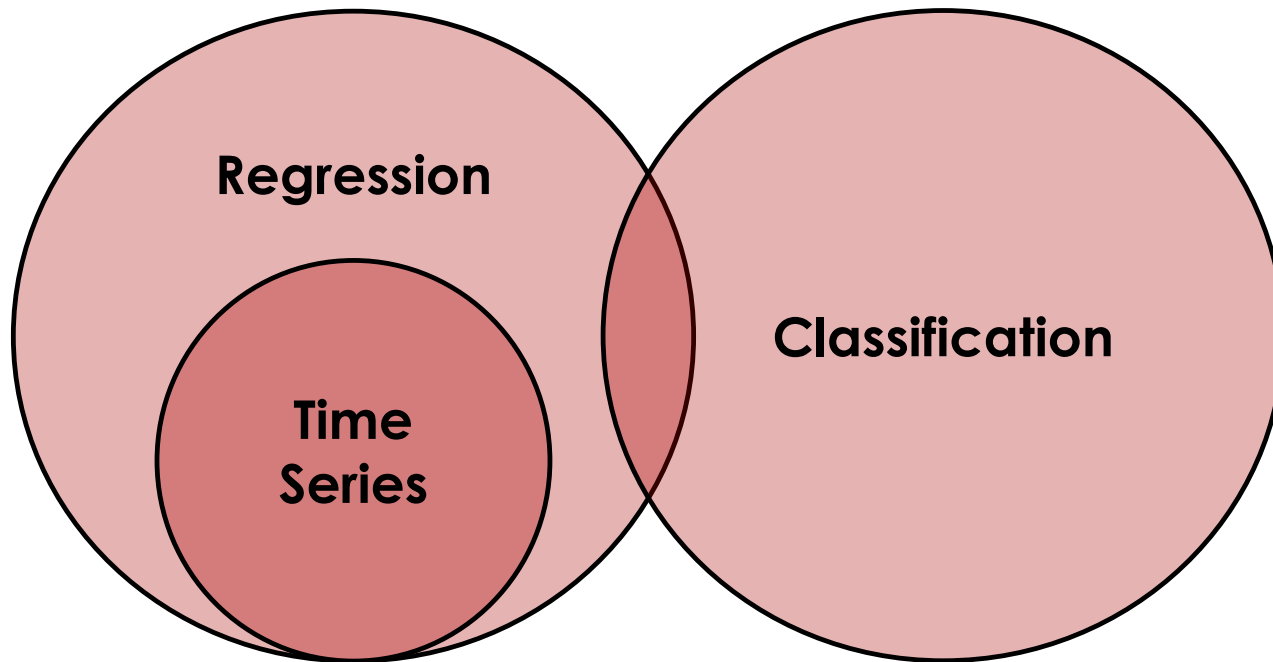
Return on the Money Market



Months since 1 Jan 1975

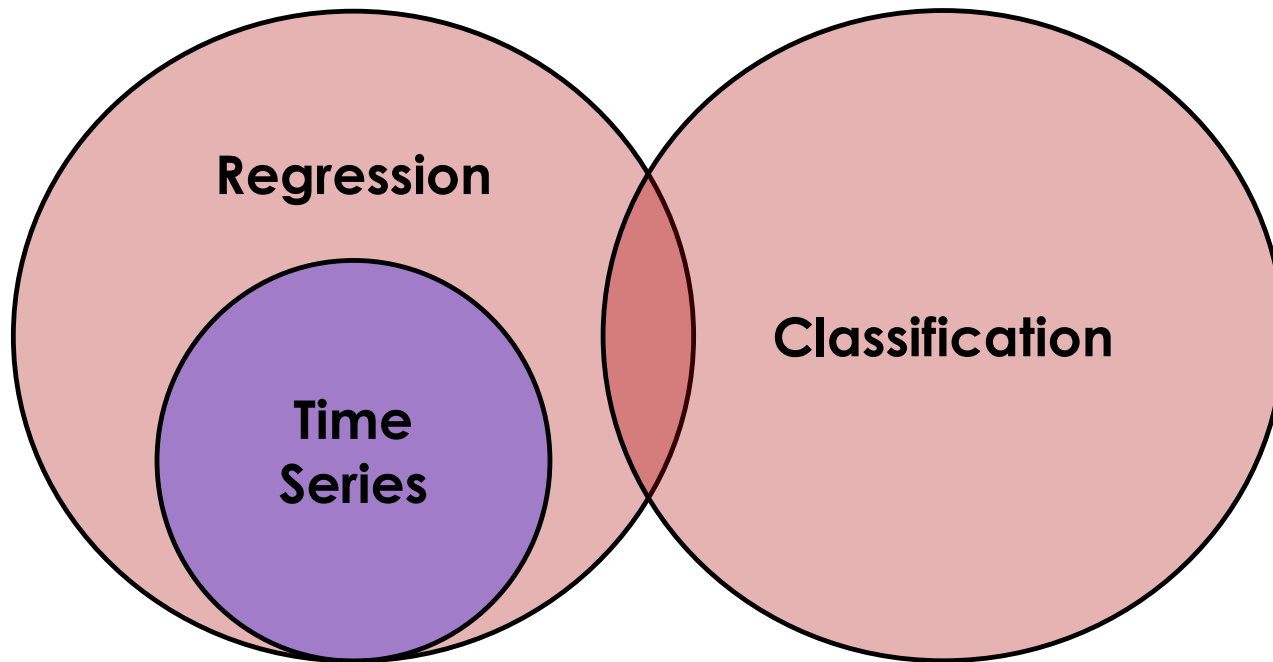
5. Problems ANNs can solve

Three types of problems:



5. Problems ANNs can solve

Three types of problems:



Time Series Problem: One Month Inflation Prediction

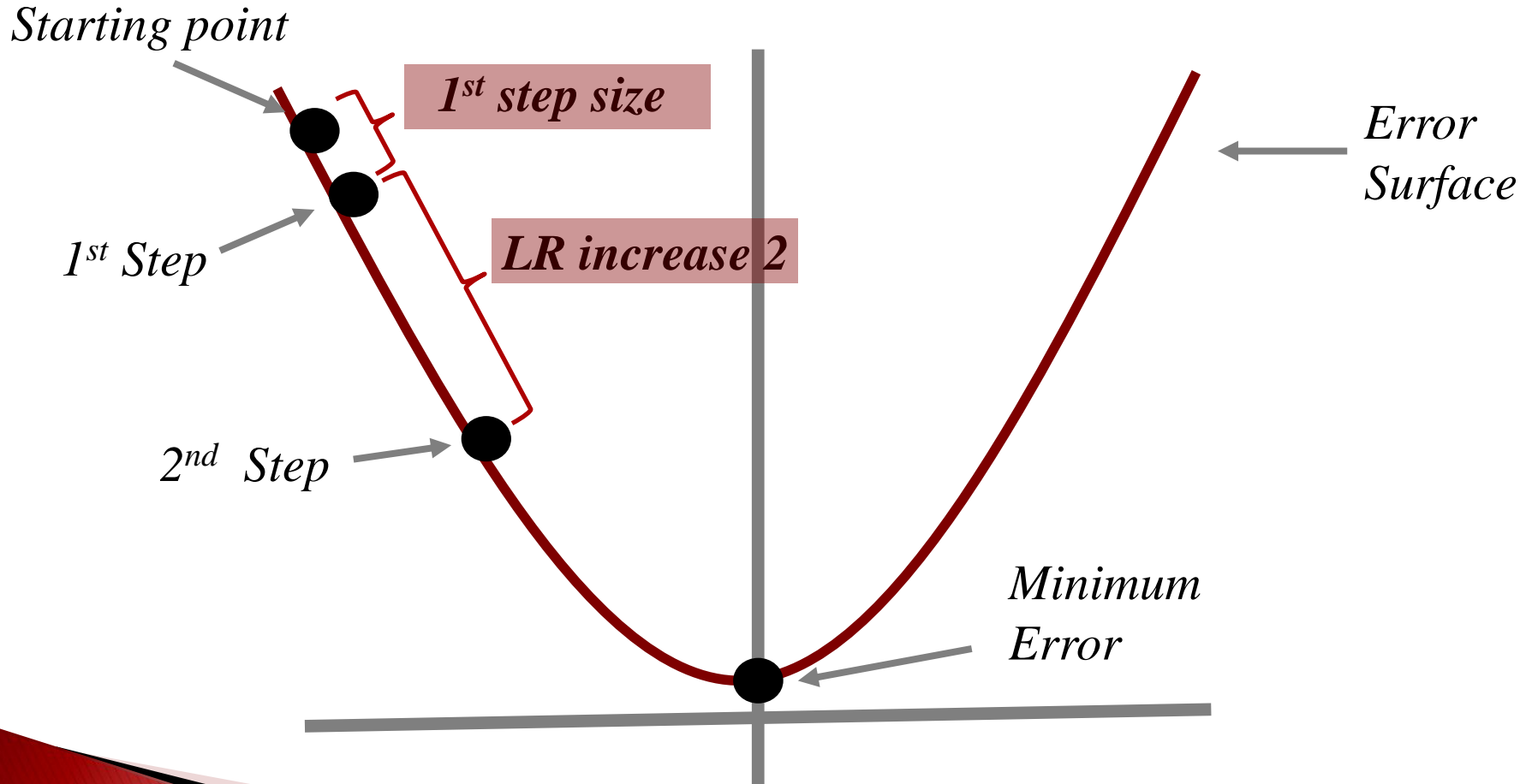
- ▶ Predict one month ahead inflation
- ▶ Sources of data:
 - Firer, C and Staunton, M., 2002. 102 Years of South African financial market history. Investment Analysts Journal, 56, 57-65
 - Firer, C. and McLeod, H., 1999. Equities, Bonds, Cash and Inflation: Historical performance in South Africa, 1925 - 1998. Investment Analysts Journal, 50, 7-28.

	Number of Observations
Training Set Size	332
Testing Set Size	100

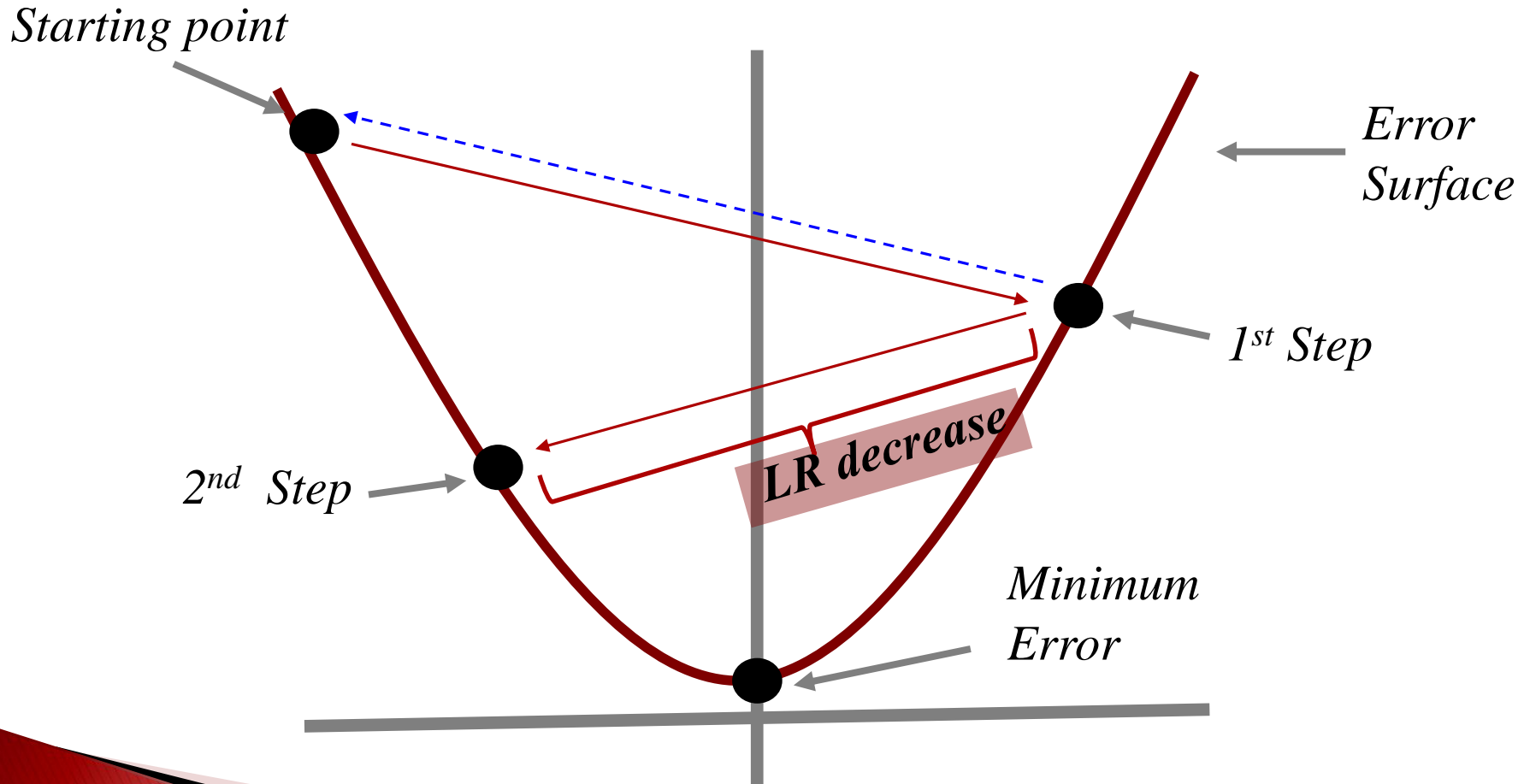
Problems

- ▶ Input Neurons
 - Partial Auto-correlations suggests 24 previous months
- ▶ Hidden Neurons
- ▶ Resilient Propagation Algorithm
 - Learning Rate increase and decrease parameters

Learning Rate Increase



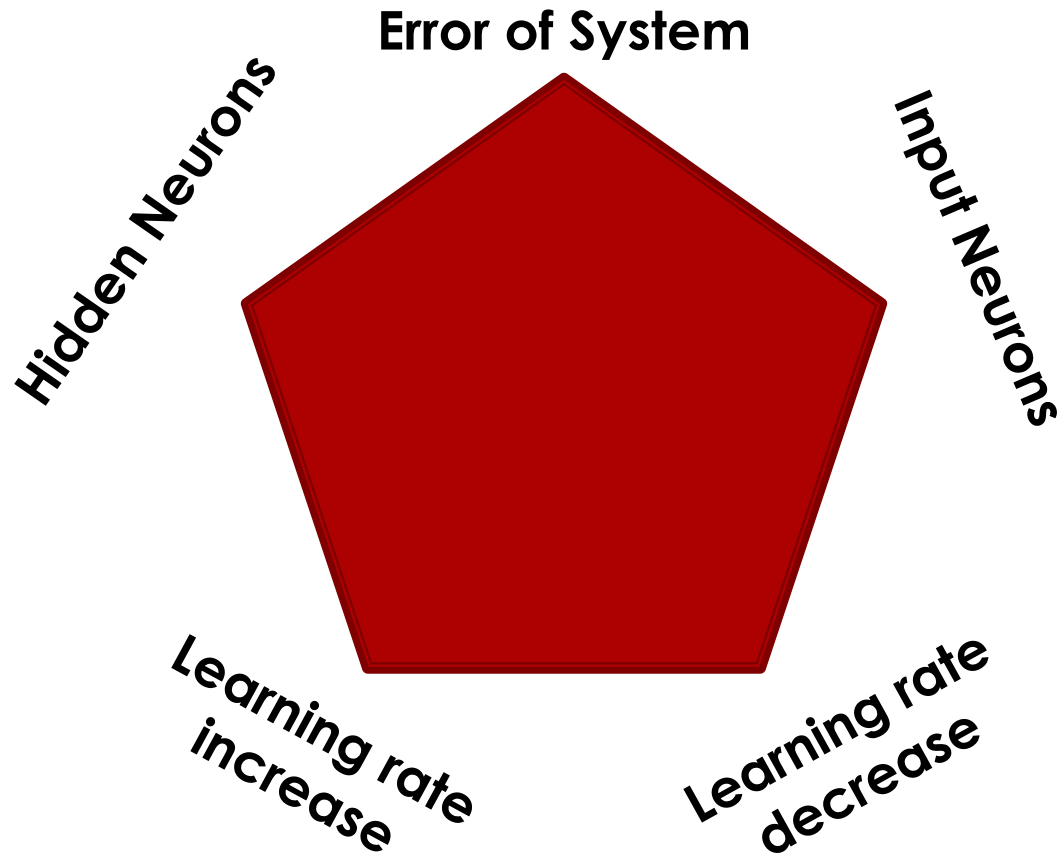
Learning Rate Decrease



Problems

- ▶ Input Neurons
 - Partial Auto-correlations suggests 24 previous months
- ▶ Hidden Neurons
- ▶ Resilient Propagation Algorithm
 - Learning Rate increase and decrease parameters

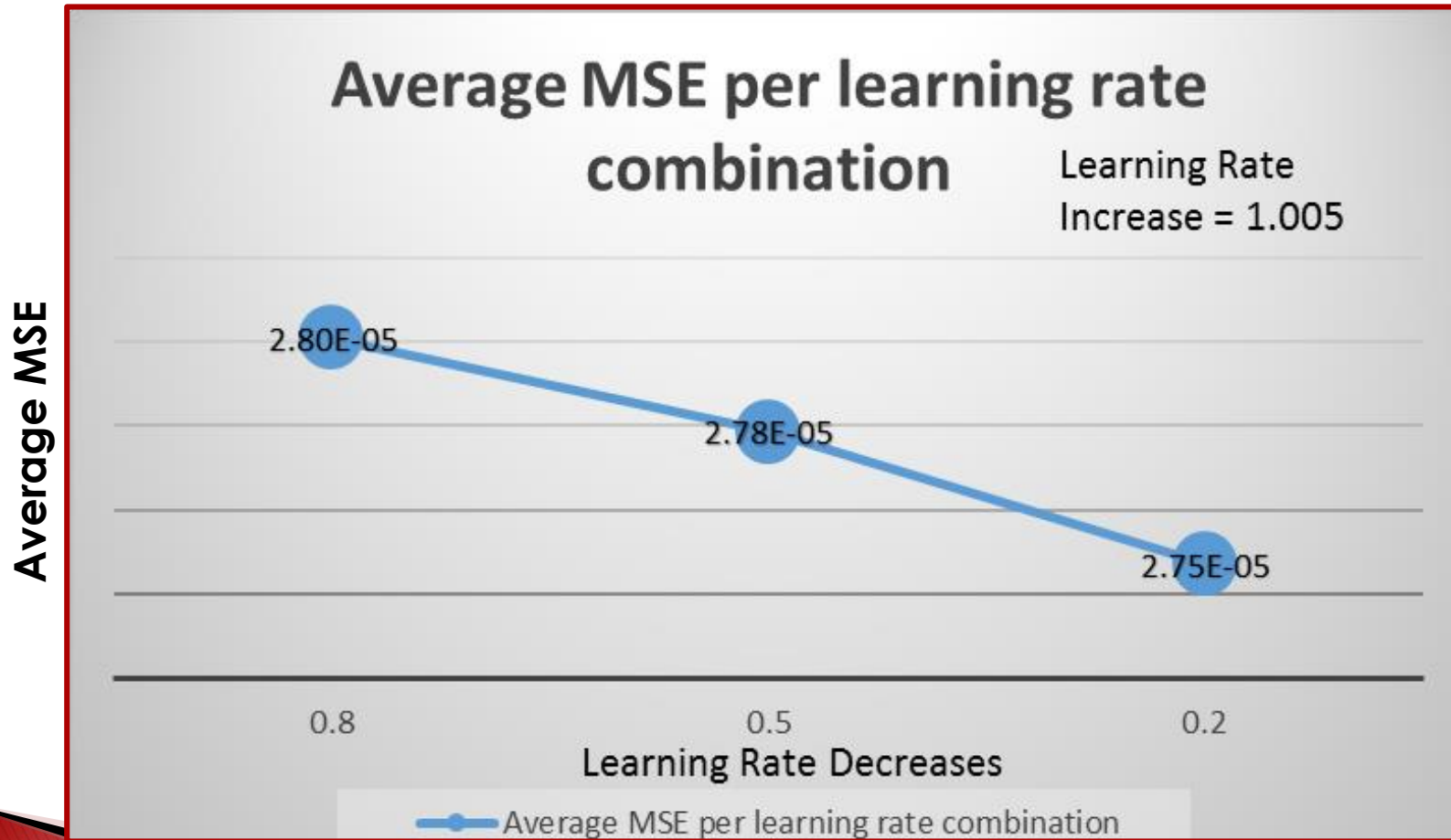
Solution



Experiment: Predicting Inflation

Learning rate changes

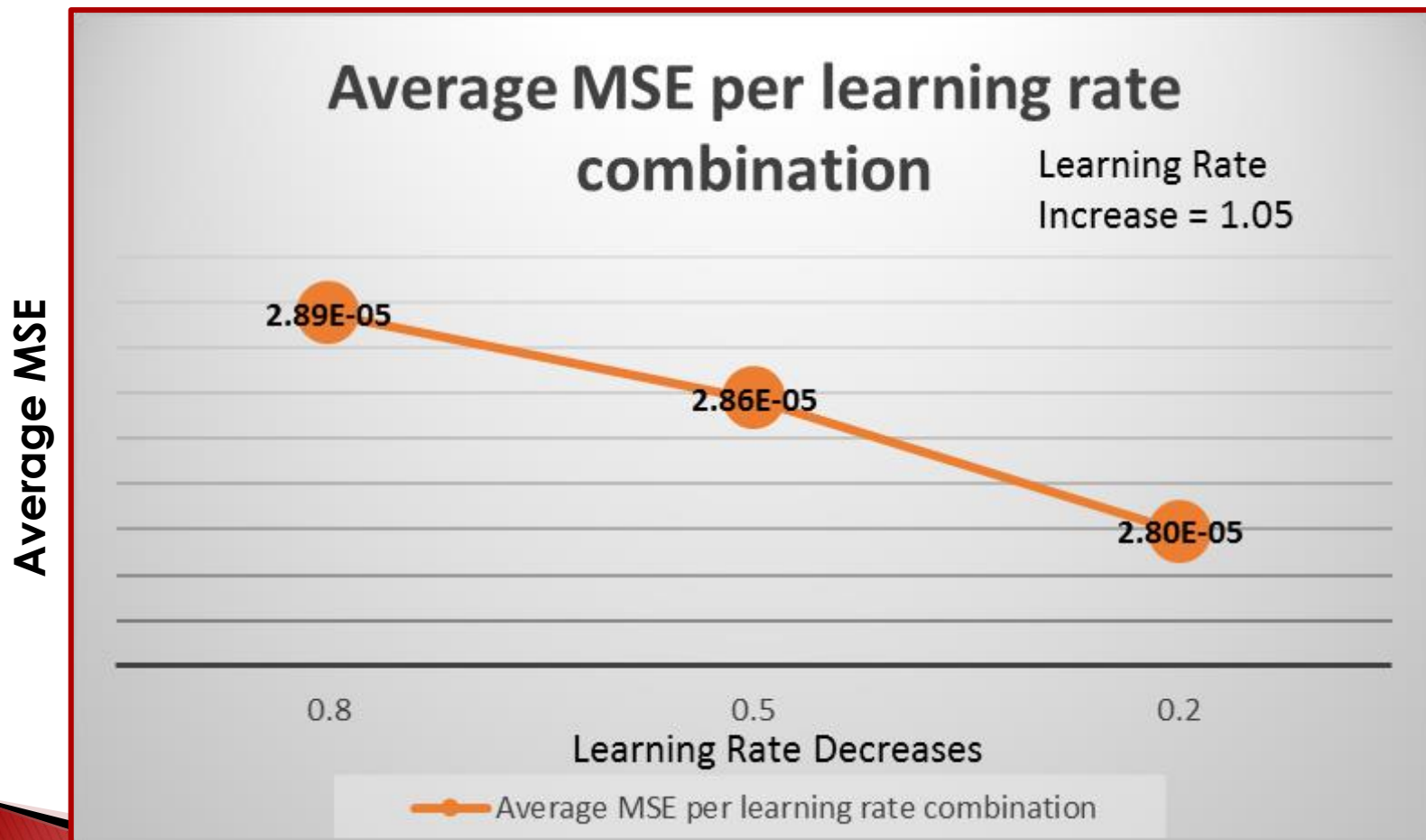
Decreases with Increase of 0.5%



Experiment: Predicting Inflation

Learning rate changes

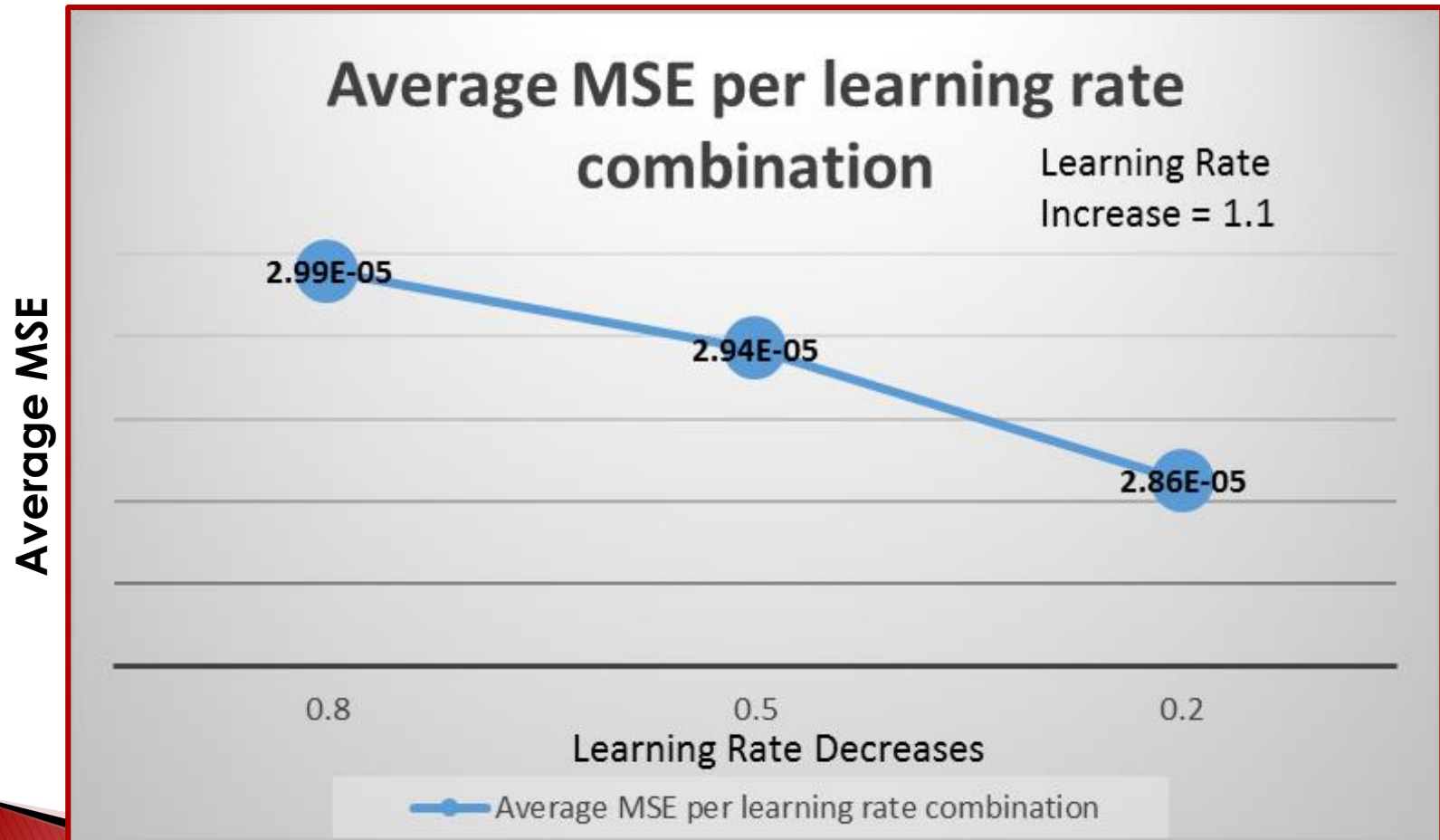
Decreases with Increase of 5%



Experiment: Predicting Inflation

Learning rate changes

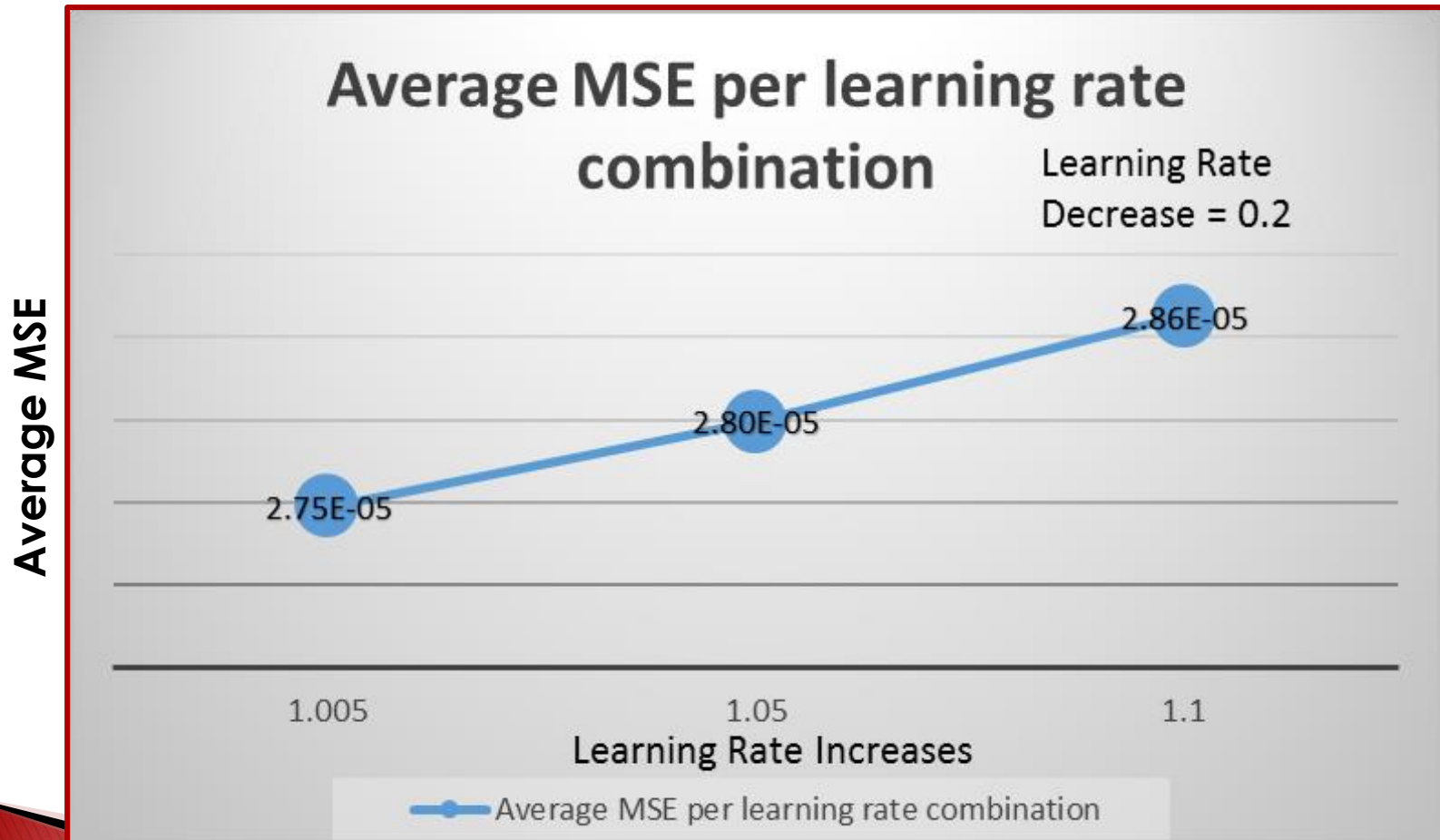
Decreases with Increase of 10%



Experiment: Predicting Inflation

Learning rate changes

Increases with Decrease of 20%



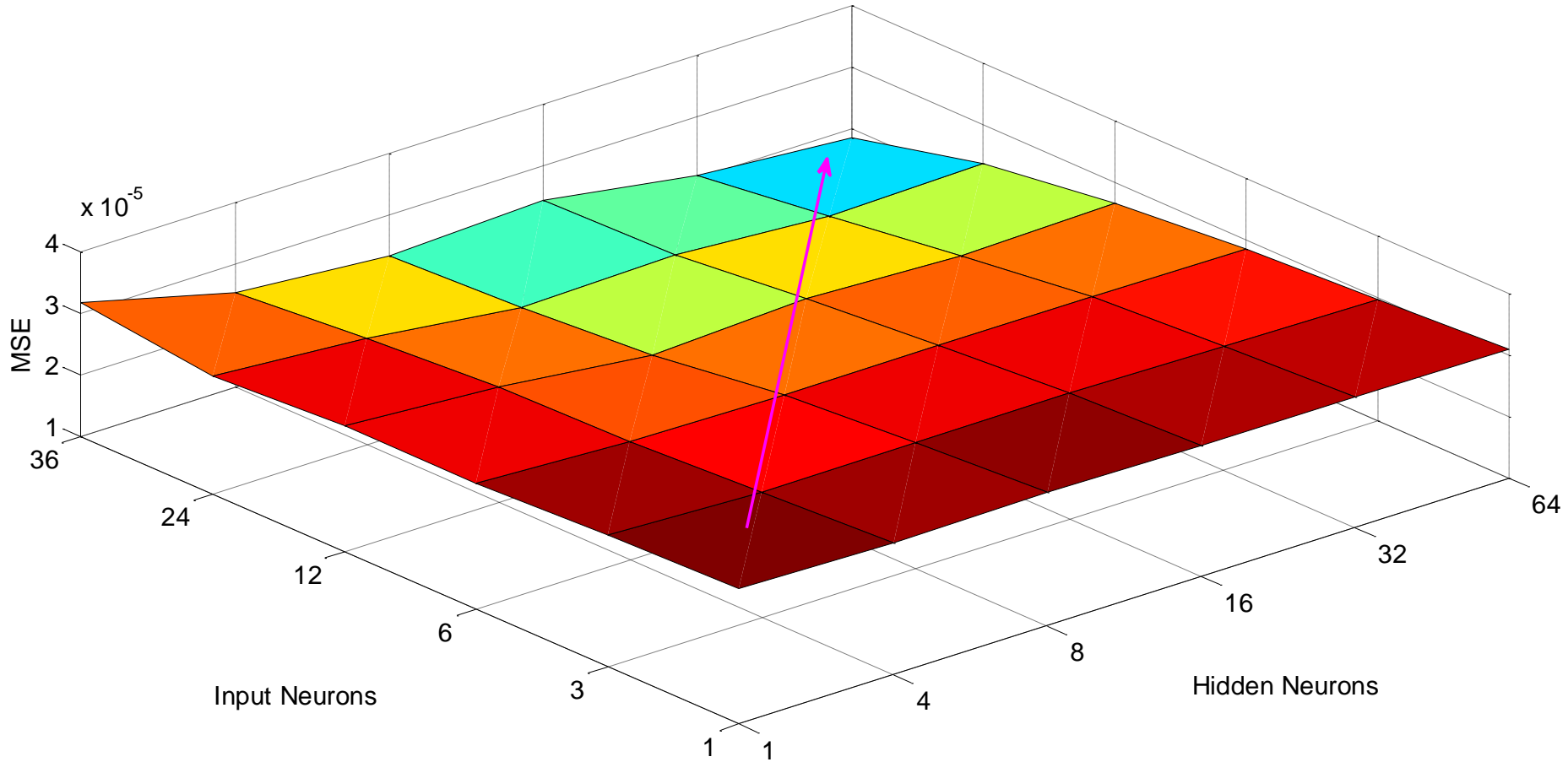
Parameters



Learning Rate increase	Learning Rate Decrease
0.5%	20%

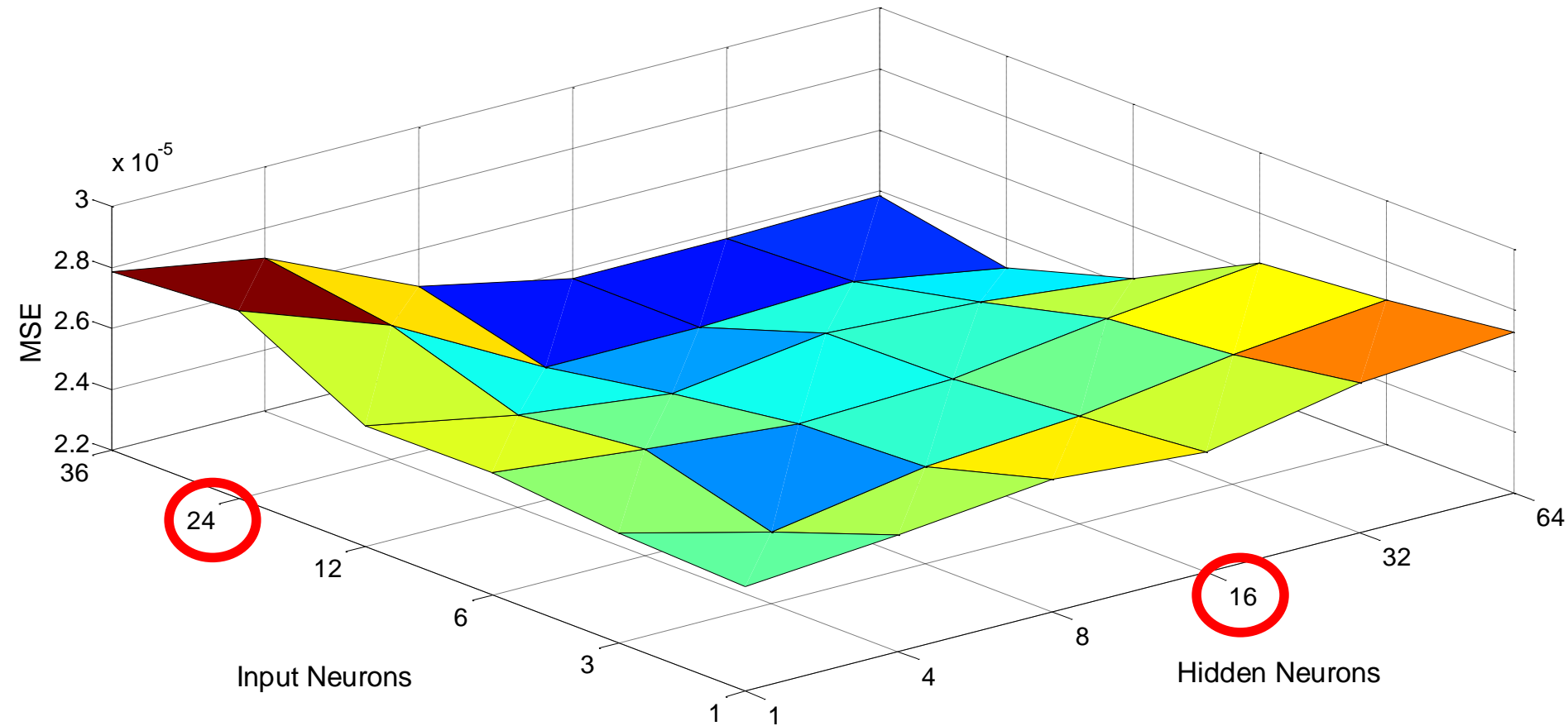
Experiment: Predicting Inflation

Best Estimate Error surface with CI of hidden vs lags - training



Experiment: Predicting Inflation

Best Estimate Error surface with CI of hidden vs lags - testing



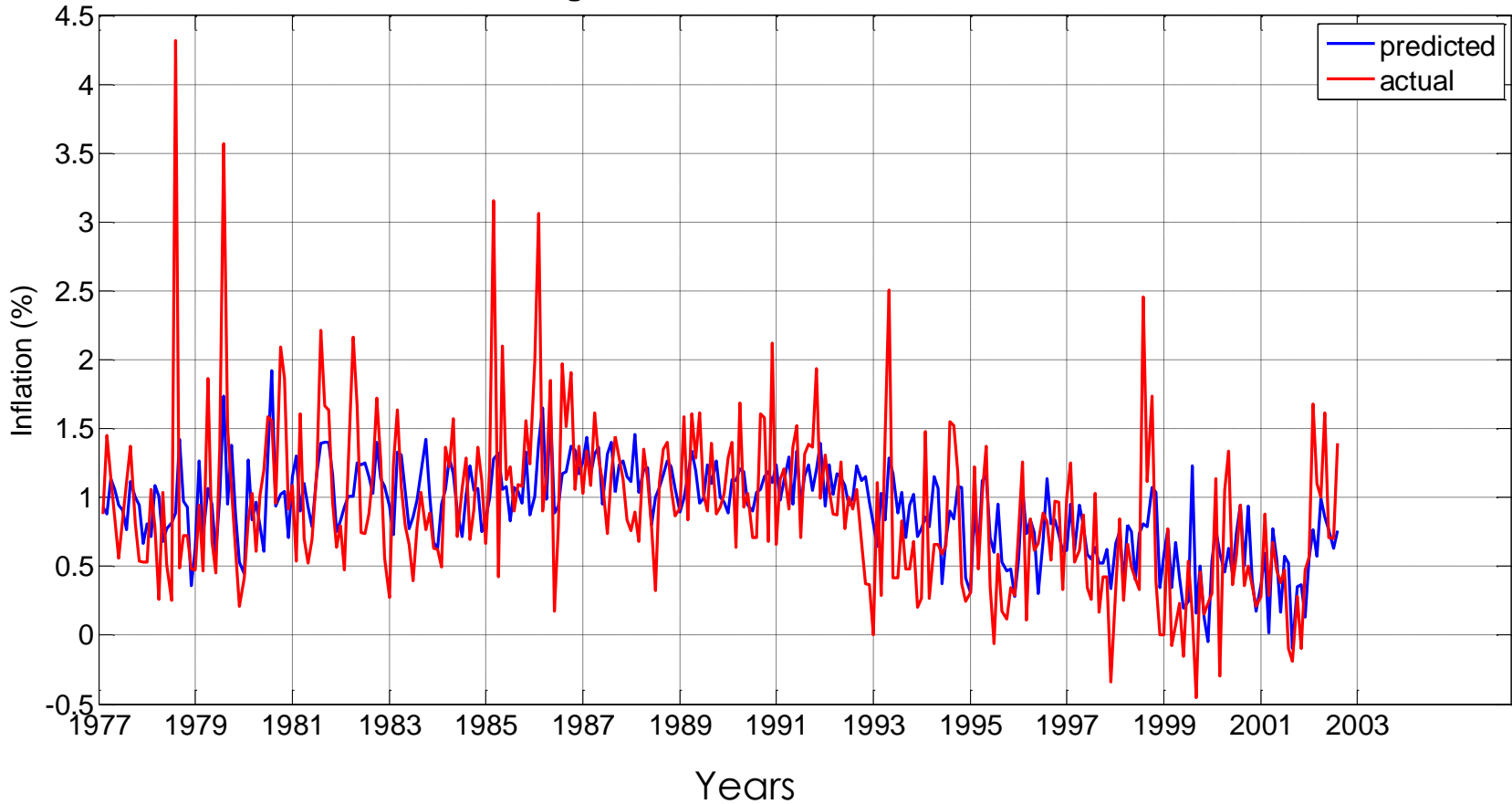
Experiment: Predicting Inflation

Optimal ANN

Structure	ANN – Training Set MSE	ANN – Training Set RMSE	ANN – Testing Set MSE	ANN – Testing Set RMSE
24-16-1	0.000022	0.0047	0.0000229	0.0048

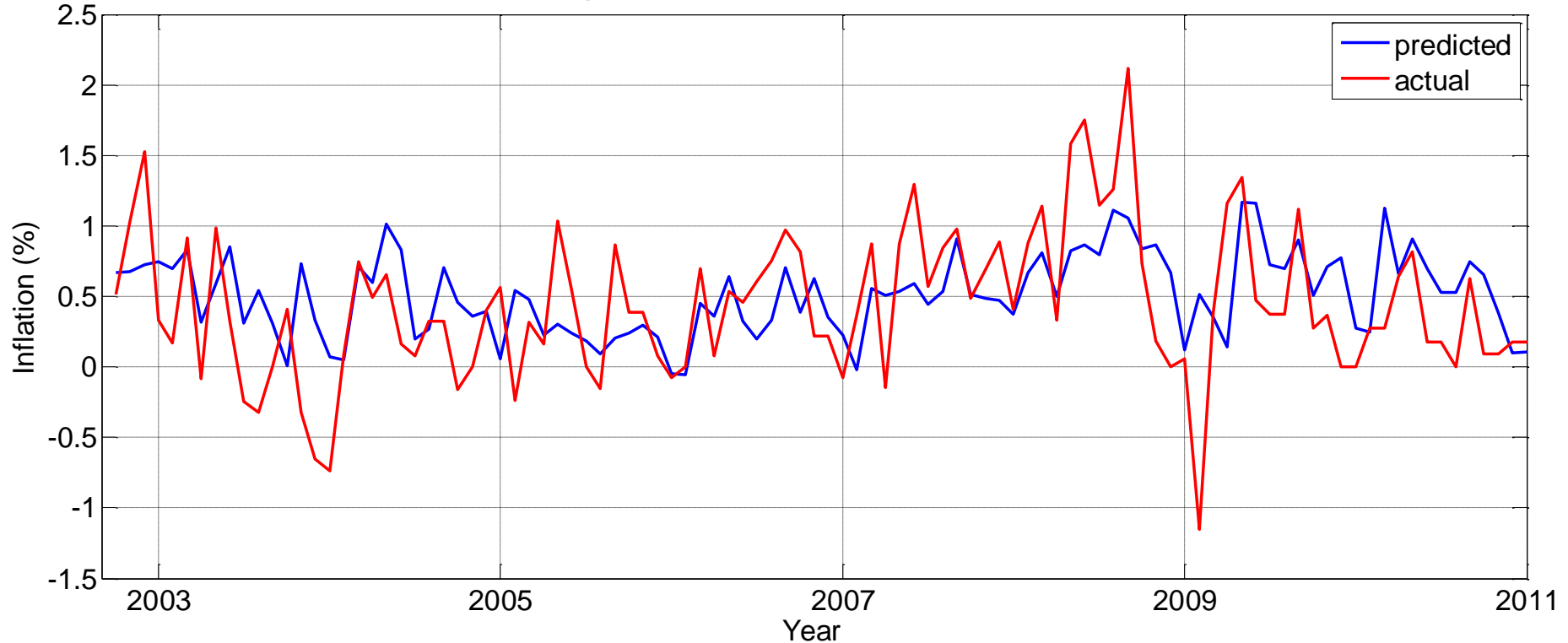
Experiment: Predicting Inflation Output

Average estimate vs actual - TRAINING SET



Experiment: Predicting Inflation Output

Average estimate vs actual - TESTING SET



Time Series Problem:

Results 1

	ANN – Training Set	ANN– Testing Data (100 data points)	Exponential Smoothing	AR(13)
Root Mean Squared Error	0.0045	0.0048	0.00535	0.00504

Problems:

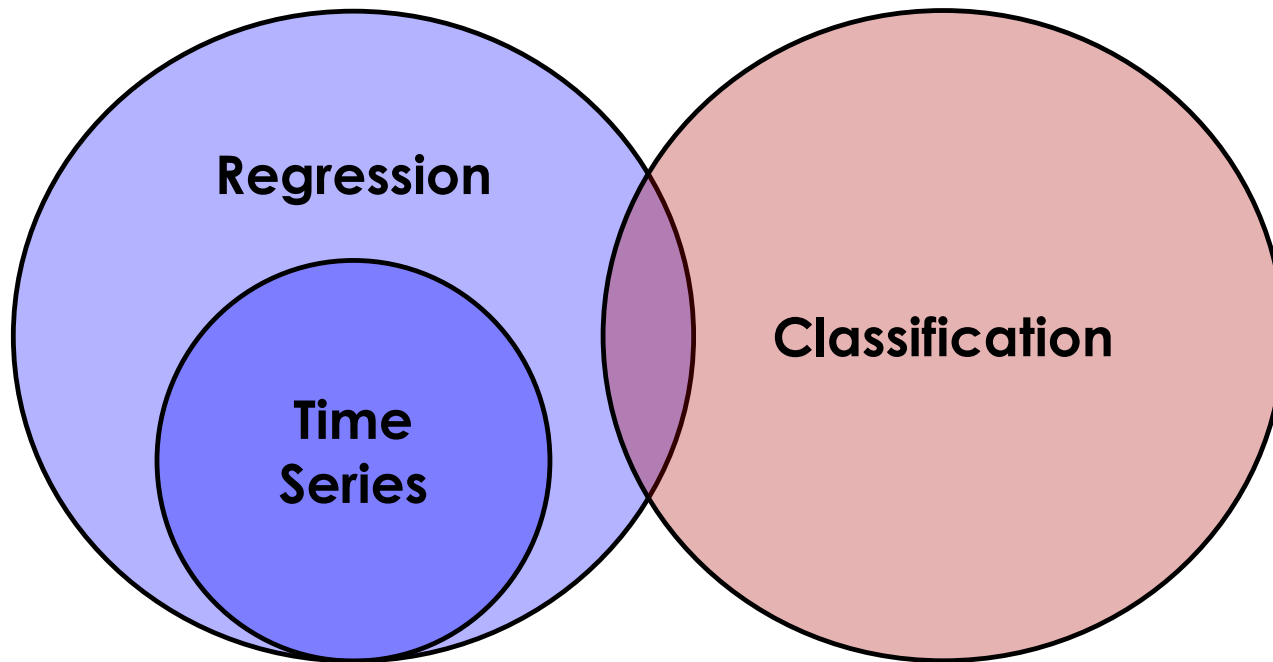
- Not consistent comparison
- Exponential and AR(13) have the advantage

Expectations:

- ANNs will perform best when considering more than 1 time step predictions

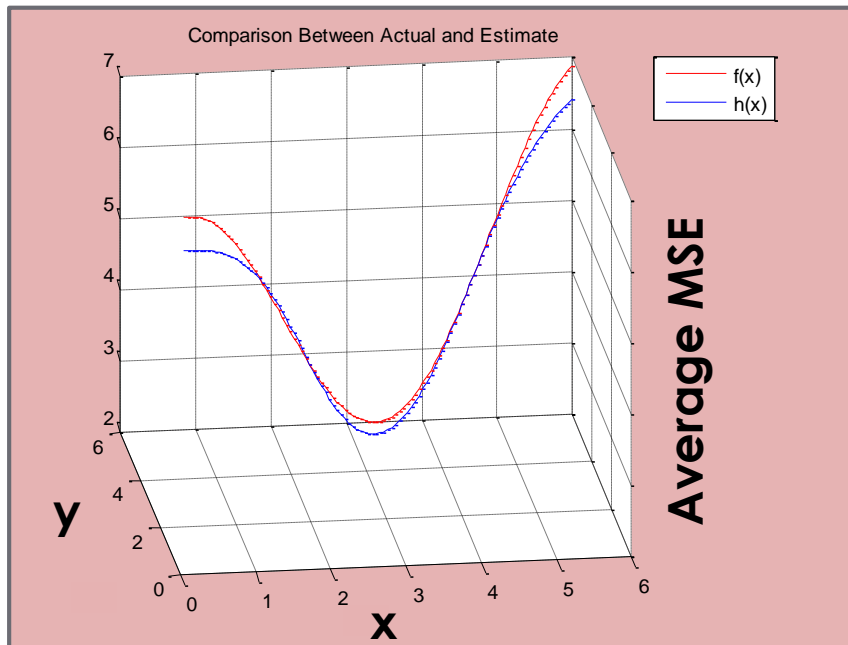
5. Problems ANNs can solve

Three types of problems:

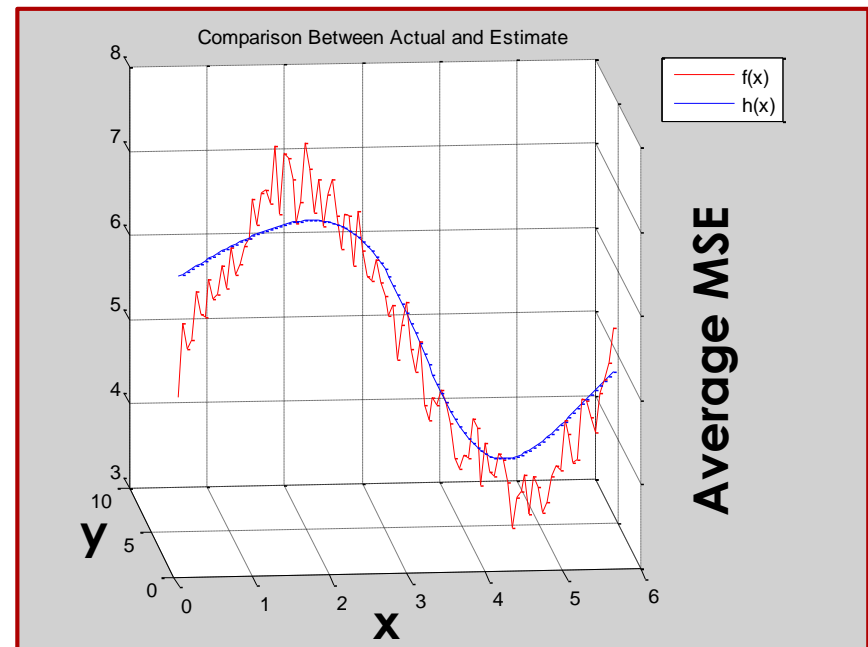


Regression Problem: Function Tracker

Cosine function

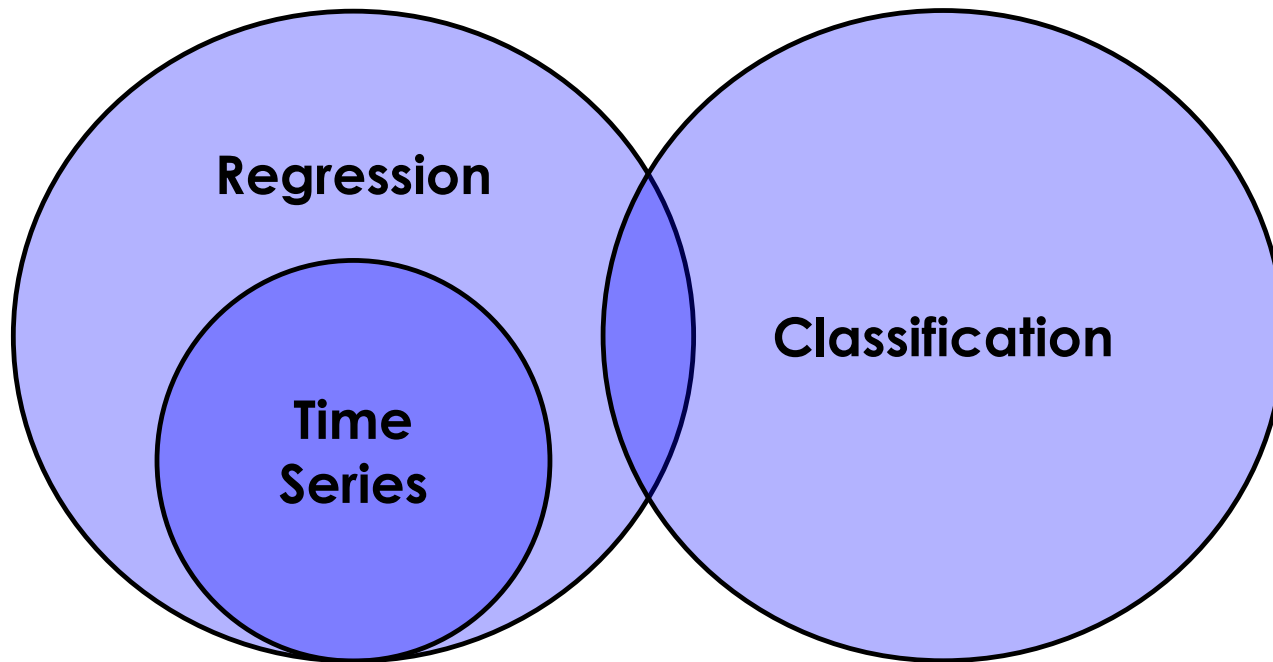


Sin function with noise



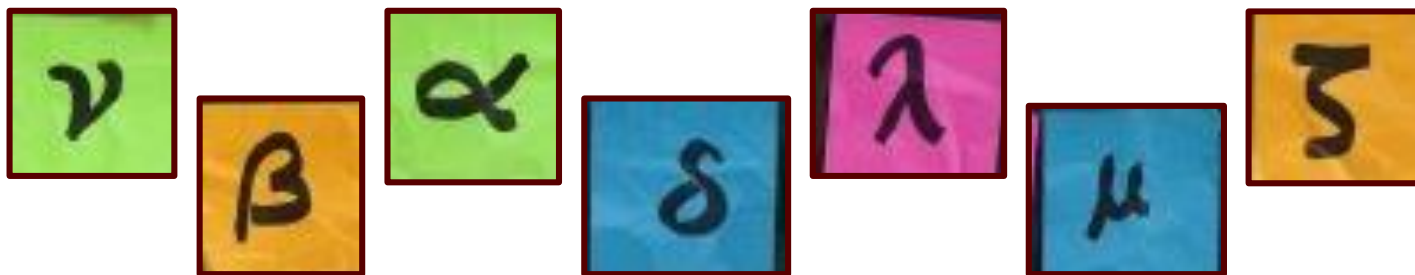
5. Problems ANNs can solve

Three types of problems:



Classification Problem: Handwriting Classification

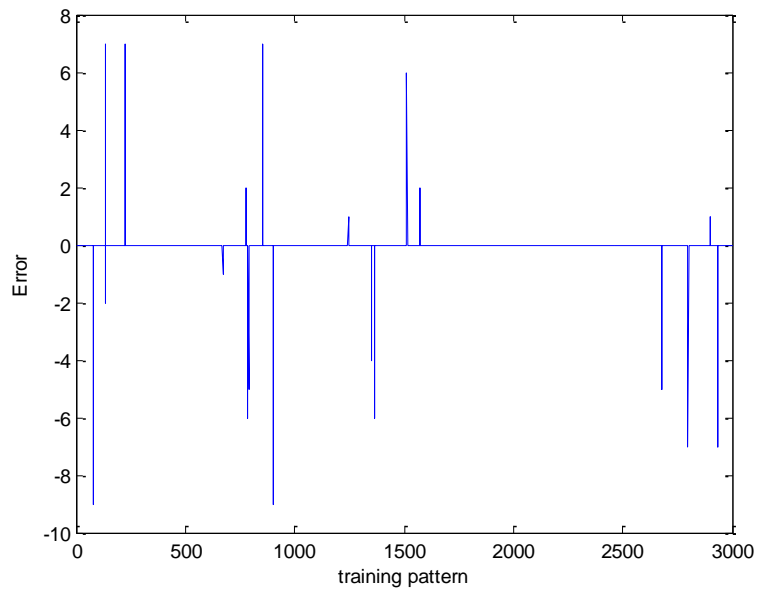
- ▶ Classify written symbols in 1 of 10 categories
- ▶ Training data – 3000 images
- ▶ Testing data – 200 images



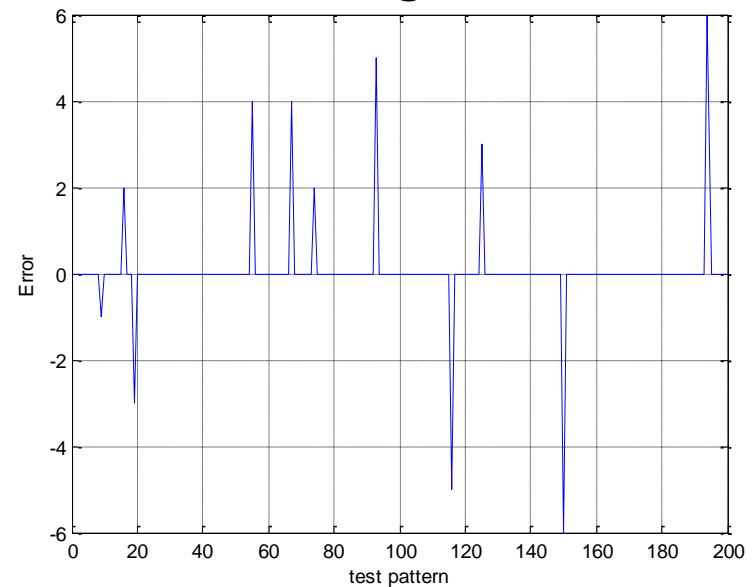
Classification Problem: Results

	Training Data	Testing Data
Classification Accuracy	99%	95%

Training Set



Testing Set



Industry Applications

- ▶ Fraud detection
 - Credit Card Fraud
 - Online Banking Fraud
 - Insurance Claims Fraud

- ▶ Assessing Mortgage Risk

- ▶ Trading Strategies
 - Buy/Sell?
 - Increase or decrease?

6. Areas of Application

- ▶ Detecting credit card fraud
- ▶ Mortgage risk assessment
- ▶ Predicting bankruptcy

Classification Applications

- ▶ Forecasting forward interest rates
- ▶ Predicting future general insurance claims
- ▶ Predicting stock exchange movements
- ▶ Forecasting commodity prices
- ▶ Forecasting foreign exchange rates

Regression and Time Series Applications

7. So why ANNs? – Summary

- ▶ Dynamic Modelling Technique
 - Detects non-linear relationships
 - Adapts to changing environments
- ▶ Types of problems:
 - Regression
 - Time Series
 - Classification



Future work

- ▶ Money Market
- ▶ Bond Market
- ▶ Equity Market
- ▶ Combination of markets
- ▶ Fraud detection



Thank you



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(Department Electronic, Electric and Computer Engineering)

- ▶ Dr C. Beyers

(Department of Insurance and Actuarial Science)

Questions?

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