

*Fast Neural Network...  
a no brainer!*

*RICCARDO TERRELL*



# Agenda

What types of problems does a neural network solve?

What exactly is a neural network?

How does a neural network actually work?

Demonstrate How to Parallelize a Neural Network

Demonstrate a Neural Network in Action

# *Introduction - Riccardo Terrell*

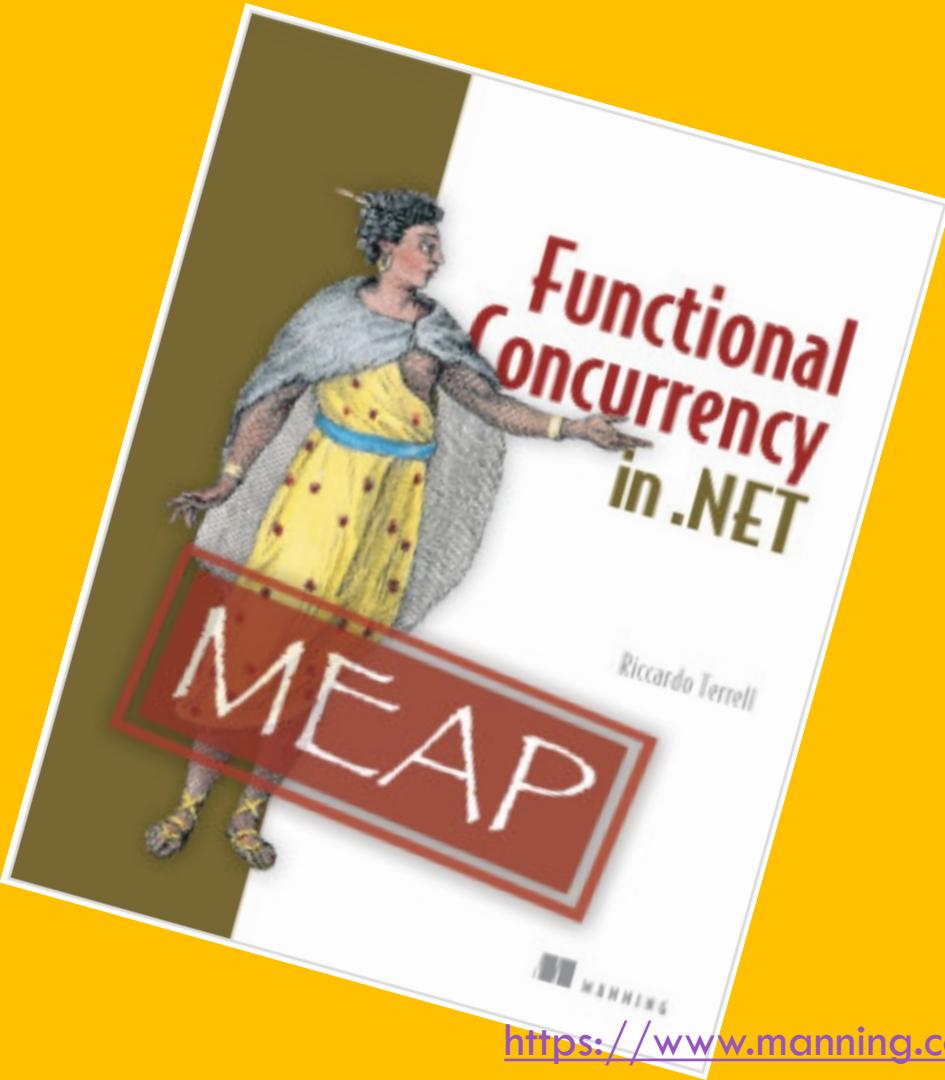
- Originally from Italy, currently - Living/working in Washington DC ~10 years
- +/- 19 years in professional programming
  - C++/VB → Java → .Net C# → Scala → Haskell → C# & F# → ??
- DC F# User Group - Organizer
- Working @  statmuse
- Polyglot programmer – I believe in the art of finding the right tool for the job



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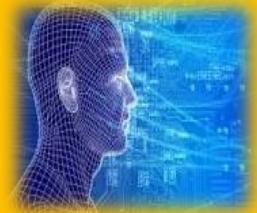
Expiration 20 February

<https://www.manning.com/books/functional-concurrency-in-dotnet>



What about you?

# Goals



Neural Networks  
ease complex  
problem  
solving...

Neural Networks  
are based on  
simple linear  
algebra!

Neural Networks  
can be  
parallelized!!  
...to be very fast



# Machine Learning Algorithms

- *Linear Regression*
- *Logistic Regression*
- *Decision Tree*
- *SVM*
- *Naive Bayes*
- *KNN*
- *K-Means*
- *Random Forest*
- *Dimensionality Reduction Algorithms*
- *Gradient Boost*
- *Adaboost*



Supervised Learning



Unsupervised Learning



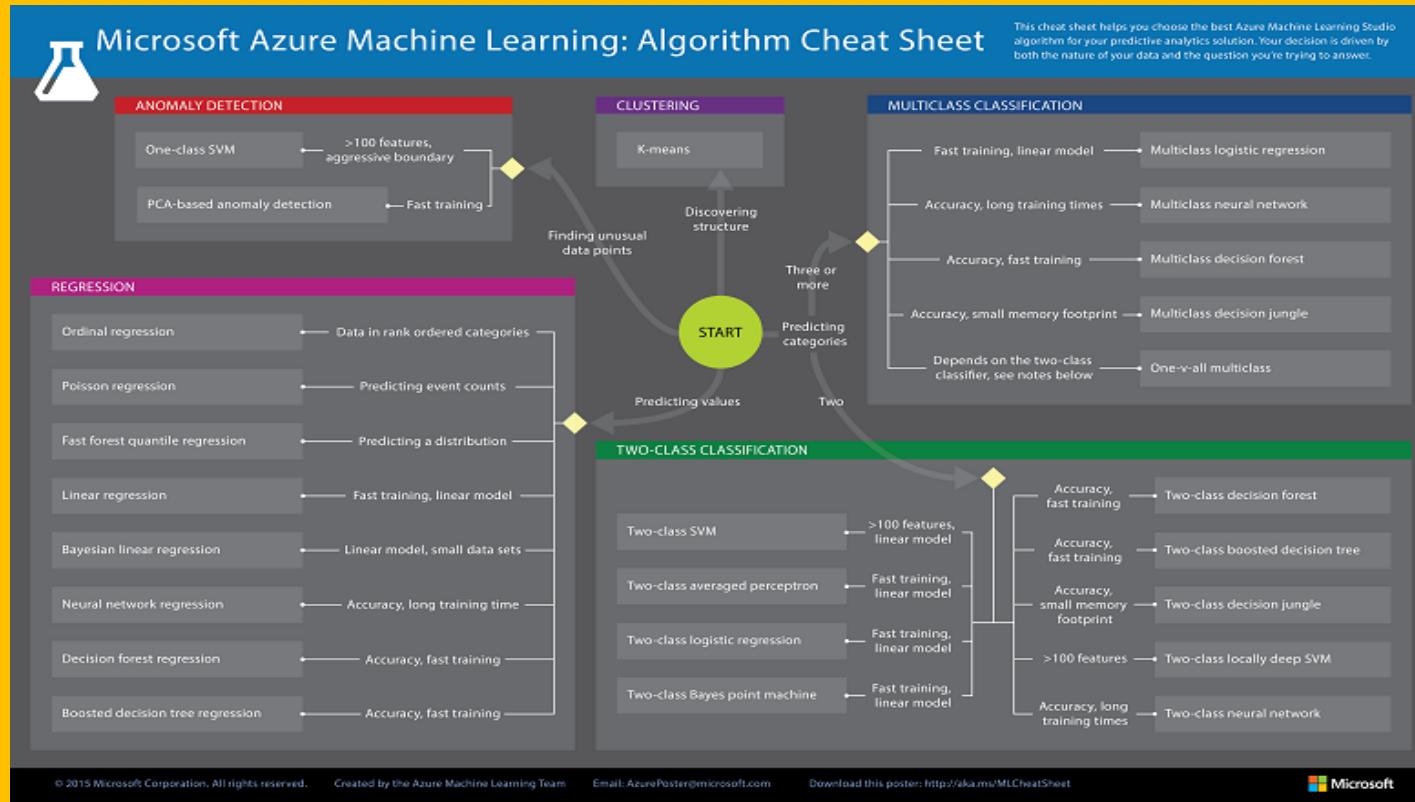
[dataaspirant.wordpress.com](http://dataaspirant.wordpress.com)

# Unsupervised & Supervised Learning

- unsupervised learning:
  - gaussian mixture models
  - latent semantic analysis (better pca for text)
  - self organizing maps
  - smarter interfaces (API to deal with reduction chaining)
  - multi-processor, gpgpu
  
- supervised learning:
  - support vector machines (non-linear classification models)
  - stochastic models (naïve bayes, etc.)
  - neural networks
  - simplified interface (system chooses appropriate model by interrogating label)
  - probabilistic graphical models
  - multi-processor, gpgpu



# Microsoft Azure ML Cheat Sheet

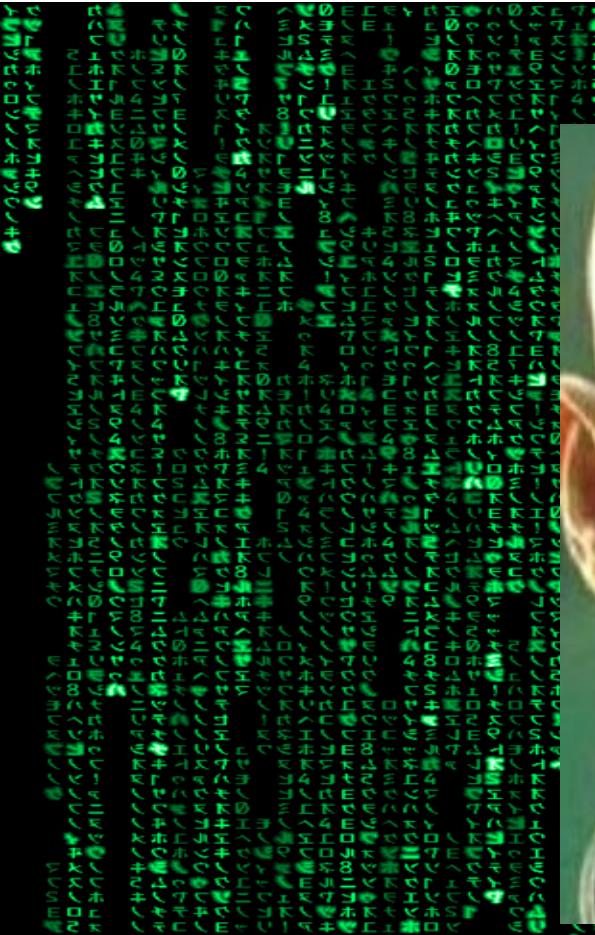


<https://docs.microsoft.com/en-us/azure/machine-learning/machine-learning-algorithm-cheat-sheet>

# WHAT IF I TOLD YOU

NEURAL NETWORK  
CAN HELP YOU

memegenerator.net



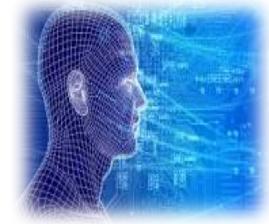
# YES!!!



NEURAL NETWORK FTW!!!!

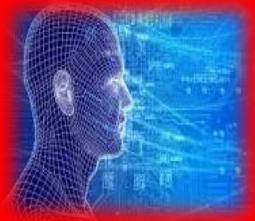
[meme-generator.net](http://meme-generator.net)

# Neural Networks are *multipurpose*



A Neural network is an alternative to:

1. *Linear regression*:
2. *Logistic regression*
3. *Naive Bayes*
4. *Decision trees*
5. *Adaptive boosting*
6. *Support vector machines*



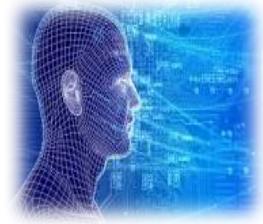
What kind of problems  
does a Neural Network  
solve?

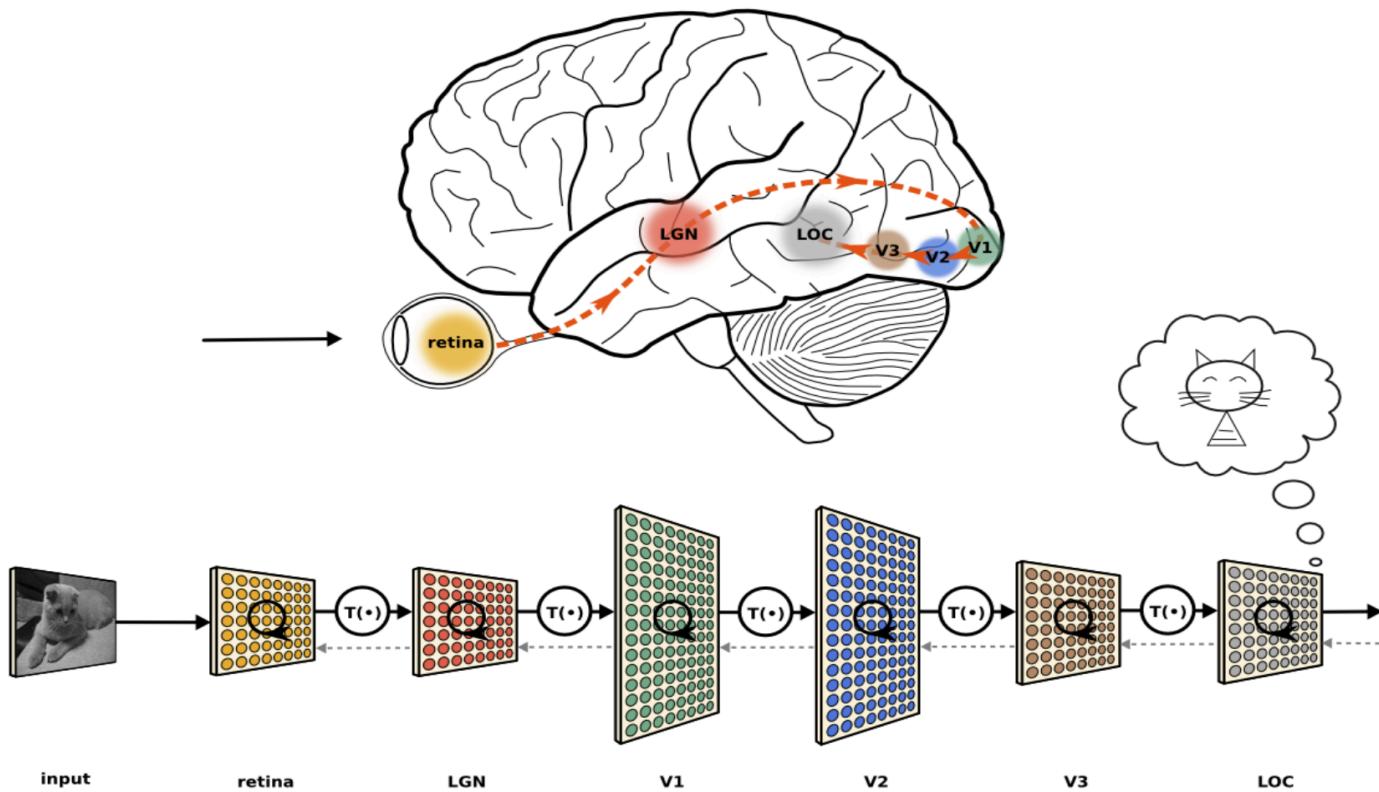
# Problems NN solve



- *OCR (optical character recognition)*
- *Natural language processing*
- *Machine translation*
- *Biology & medicine*
- *Robotics (Autonomous Systems)*
- *Pattern Recognition Time Series Prediction*
- *Signal Processing*
- *Control*
- *Soft Sensors*
- *Anomaly Detection*

# What is this?





# NN Prediction Example



Grade	GPA	Age	Gender	Sport
A	3.5	16	M	Baseball
C	2.0	12	M	Football
F	2.1	13	F	Soccer
B	3.5	17	M	Baseball
D	2.0	18	F	Soccer
A	3.8	15	M	Baseball
D	2.3	14	F	Baseball
B	3.3	17	M	????

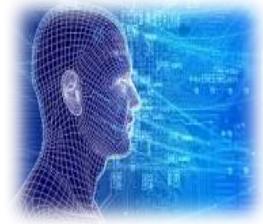
Training data

**features**  
gpa, age, gender, sport  
**values (x)**  
[B, 3.3, 17, M, Baseball]

Independent variables/predictors/  
attributes/regressors/x-values

"The thing to classify (predict)"/  
dependent variable/y

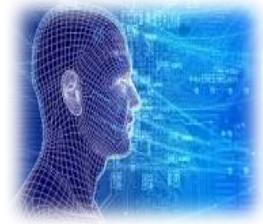
# Recommender systems



- what do people like?



# Recommender systems



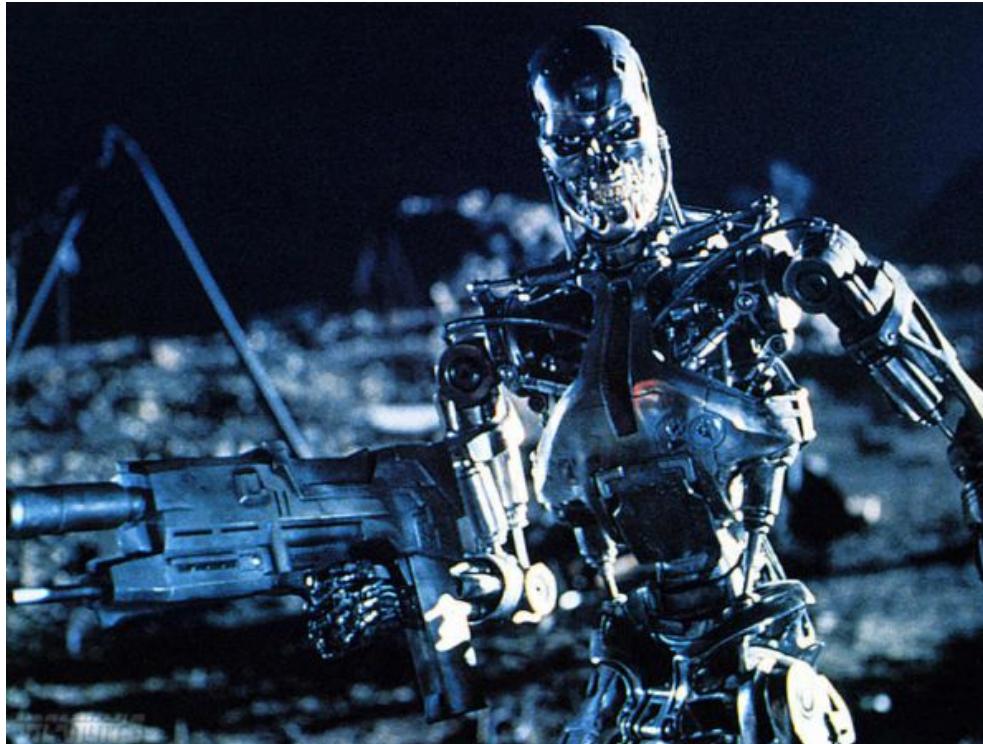
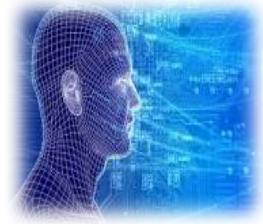
- how are things alike?



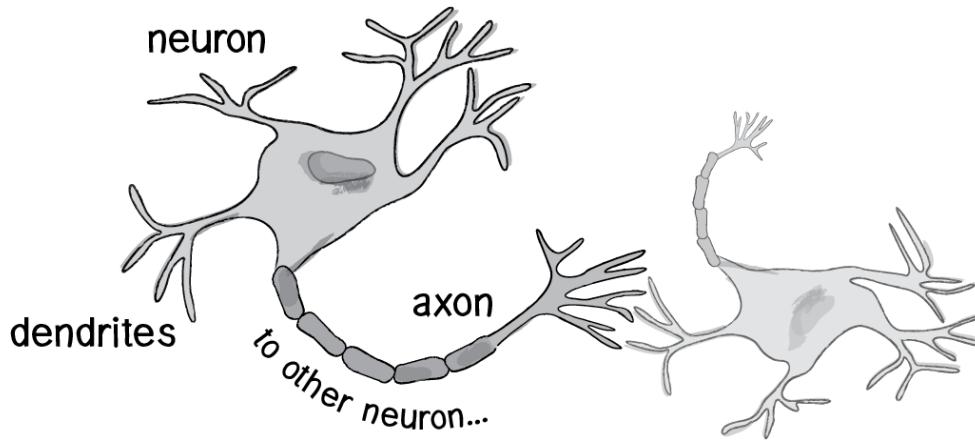


# What is a Neural Network

# What is this Neural Network?



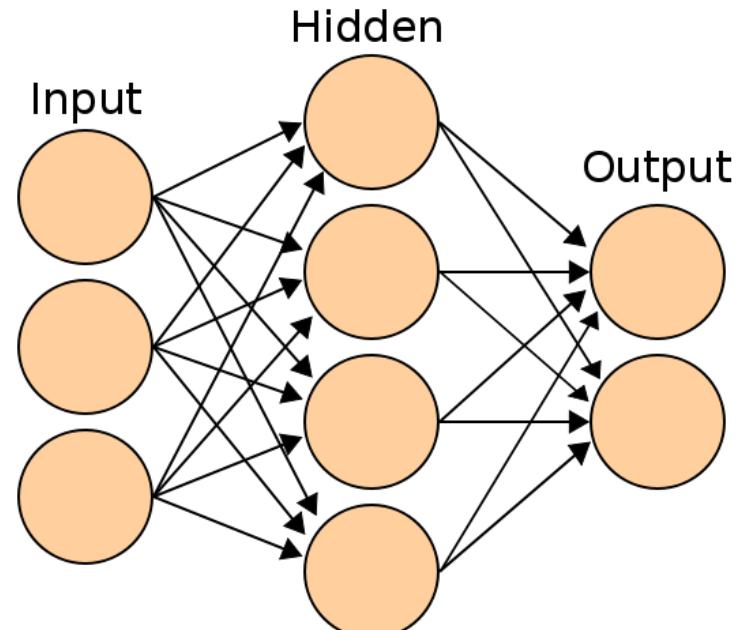
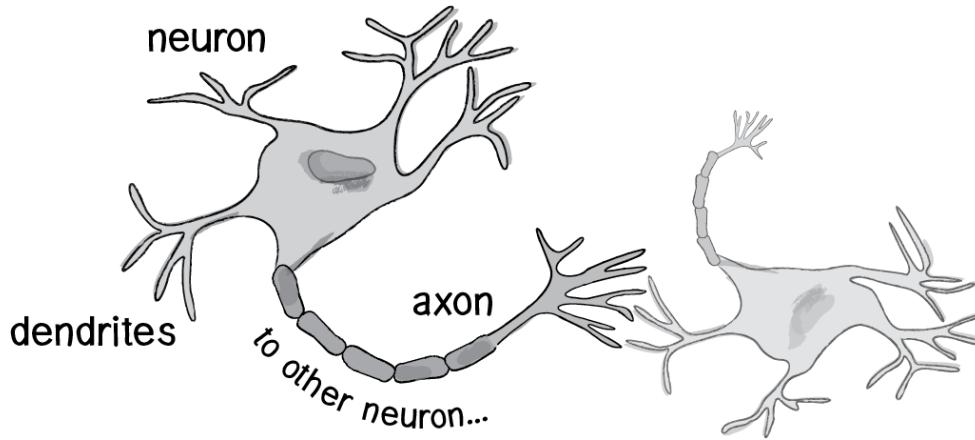
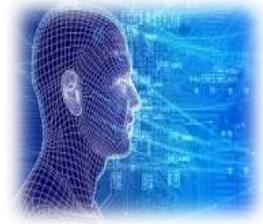
# What is Neural Network



## Four parts of a typical nerve cell :

- **DENDRITES:** Accepts the inputs
- **SOMA :** Process the inputs
- **AXON :** Turns the processed inputs into outputs.
- **SYNAPSES :** The electrochemical contact between the neurons.

# What is Neural Network



- A neuron: many-inputs / one-output unit
- output can be **excited** or not **excited**
- incoming signals from other neurons determine if the neuron shall **excite** ("fire")
- Output subject to attenuation in the **synapses**, which are junction parts of the neuron

# Mathematical Background



$$\begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 2 & 5 \end{bmatrix}$$

**Matrix Operations**

$$\sum_{i=1}^n c = nc$$

**Sigma Notation**

$$\frac{d}{dx} x^2 = 2x$$

**Equation for derivatives and Integrals**

# Some vocabulary

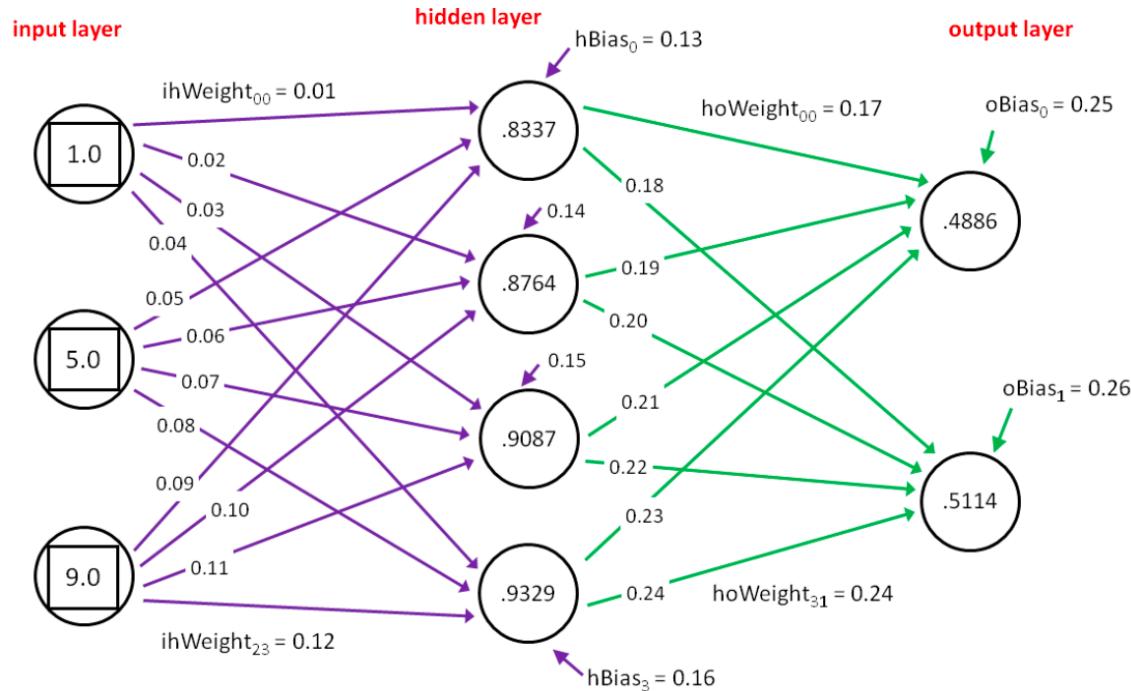
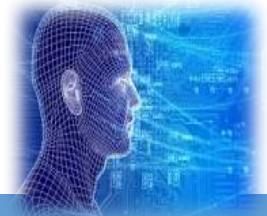


Biological Terminology	Artificial Neural Network Terminology
Neuron	Node/Unit/Cell/Neurode
Synapse	Connection/Edge/Link
Synaptic Efficiency	Connection Strength/Weight
Firing frequency	Node output

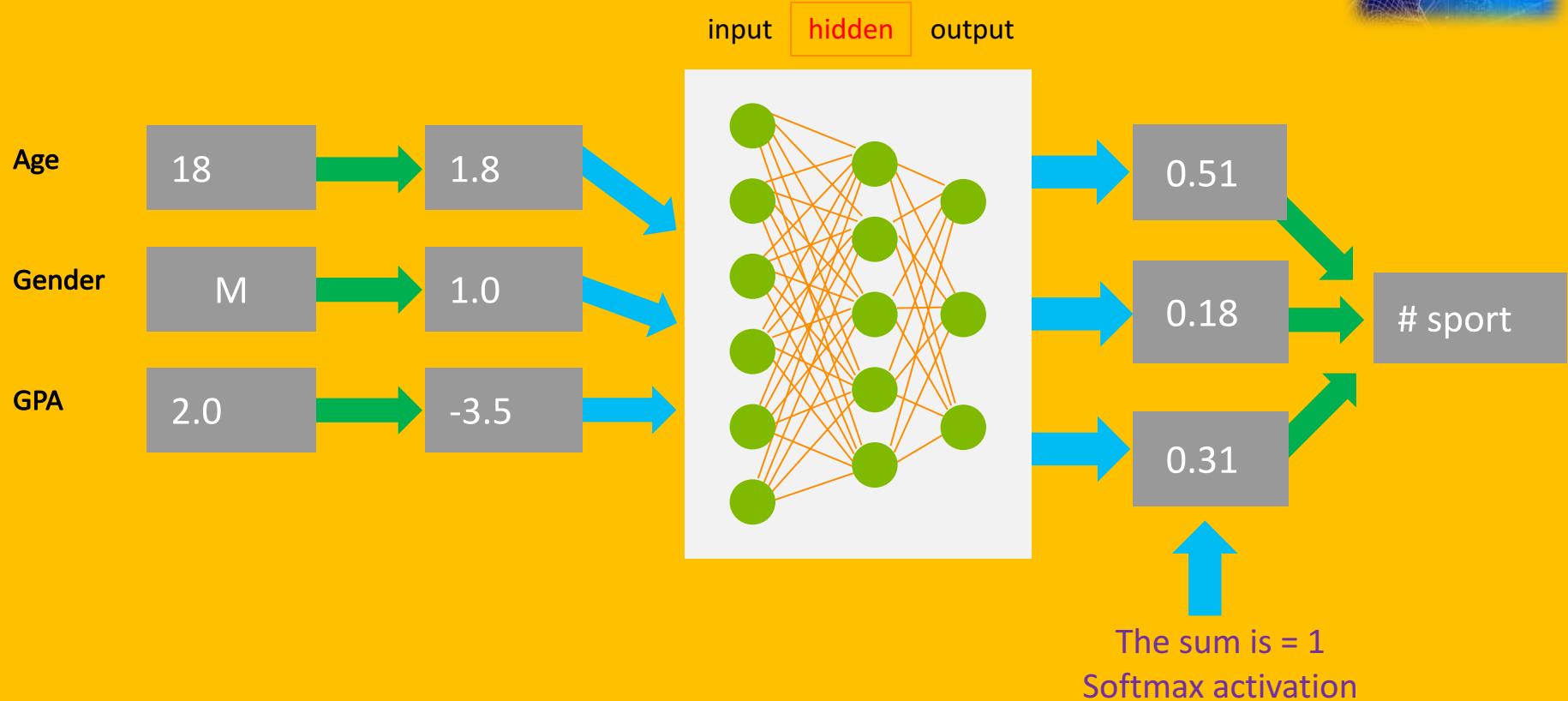


How does a  
Neural Network  
work?

# Neural Network Classification

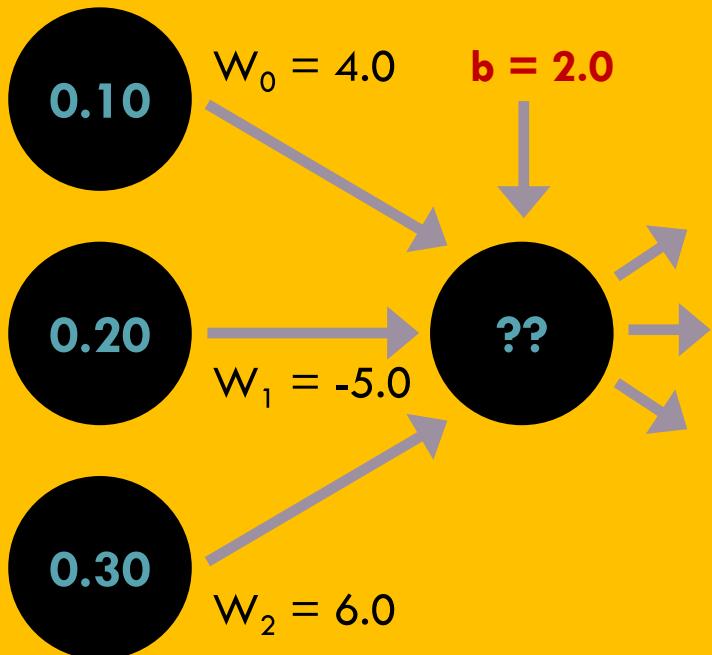


# What's going one?





# The perceptron



- a.  $(0.1)(4.0) + (0.2)(-5.0) + (0.3)(6.0) = 1.2$
- b.  $1.2 + 2.0 = 3.2$
- c. Activation Function (3.2) = 0.73
- d. ?? = 0.73

# Activation functions



## □ Logistic sigmoid

- Output between [0, 1]
- $y = 1.0 / (1.0 + e^{-x})$

## □ Hyperbolic tangent

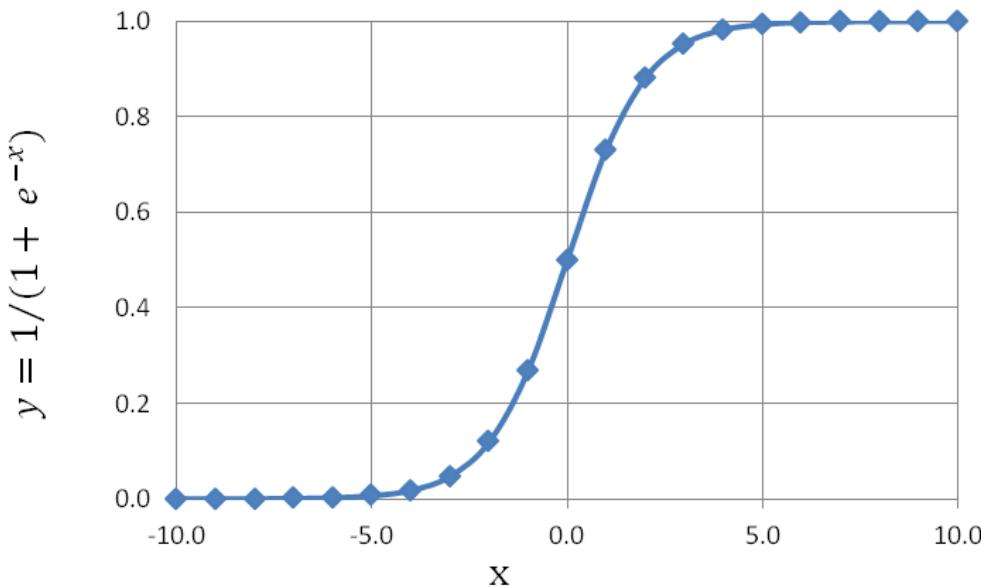
- Output between [-1, +1]
- $y = \tanh(x) = (e^x - e^{-x}) / (e^x + e^{-x})$

## □ Heaviside step

- Output either 0 or 1
- if  $(x < 0)$  then  $y = 0$  else if  $(x \geq 0)$  then  $y$

## □ Softmax

- Outputs between [0, 1] and sum to 1.0
- $y = (e^{-xi}) / \sum (e^{-xi})$



# Neural Network Training



## □ Back-propagation

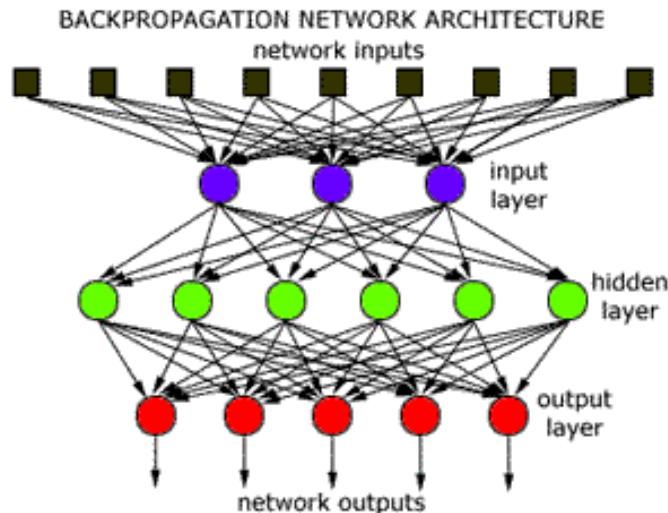
- Fastest technique.
- Does not work with Heaviside activation.
- Requires “learning rate” and “momentum.”

## □ Genetic algorithm

- Slowest technique.
- Generally most effective.
- Requires “population size,” “mutation rate,” “max generations,” “selection probability.”

## □ Particle swarm optimization

- Good compromise.
- Requires “number particles,” “max iterations,” “cognitive weight,” “social weight.”



**SHOW**



# Neural Network in code



```
let prepend value (vec:Vector<float>) =
    vector [ yield! value :: (vec |> Vector.toList) ]  
  
let prependForBias : Vector<float> -> Vector<float> = prepend 1.0  
  
let layer activationF (weights:Matrix<float>) (inputs:Vector<float>) =
    (weights * inputs) |> Vector.map activationF |> prependForBias
```

# Neural Network in code



```
let sigmoid x = 1.0 / (1.0 + exp(-x))

let wmd = matrix [ [-1.0; 0.1; 0.3; 0.7];
                   [-2.0; 0.3; 0.2; 1.0];
                   [-1.0; 0.8; 0.6; 0.6];
                   [-2.0; 0.6; 0.3; 0.5]; ]

let wkm = matrix [ [-1.5; 0.5; 0.5; 0.5; 0.5];
                   [-2.5; 0.5; 0.5; 0.5; 0.5]; ]

let twoLayerNetwork : (Vector<float> -> Vector<float>) =
    layer sigmoid wmd >> layer sigmoid wkm

let compute (input : Vector<float>) =
    (twoLayerNetwork <| prependForBias input).[ 1 .. ]
```

# Neural Network in code



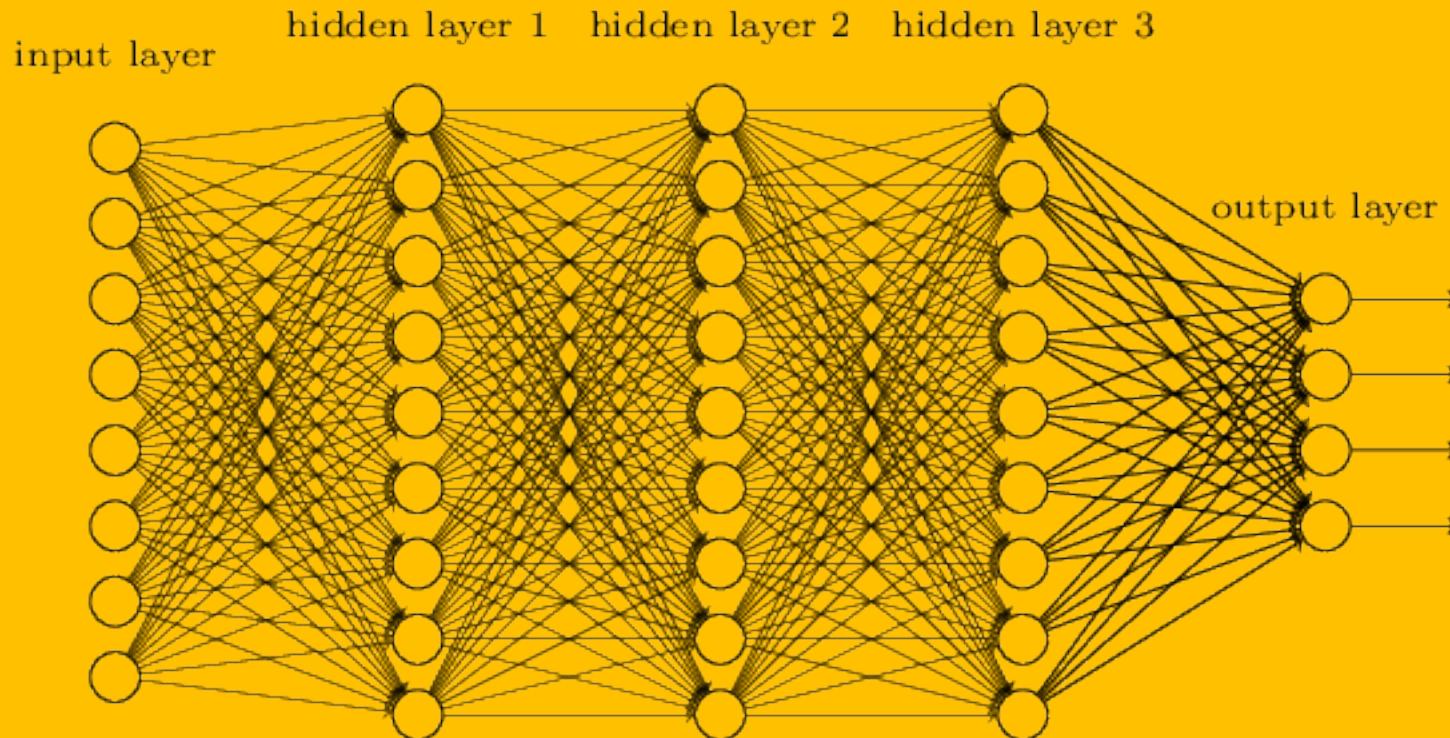
```
let out = compute (vector [1.0; 2.0; 3.0])
```

```
RESULT ➔ Vector<float> = seq [0.5392375632; 0.3009608905]
```

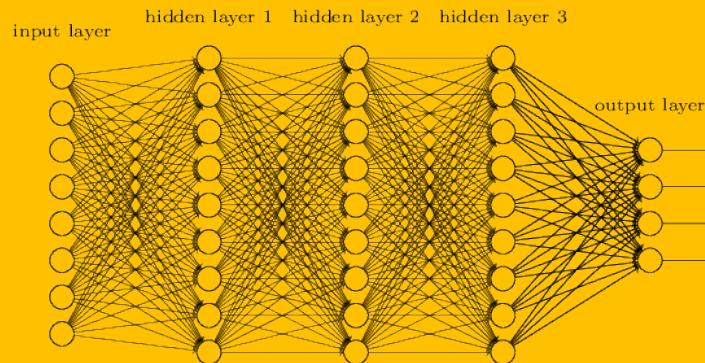


# Why a Parallel Neural Network?

# Why a Parallel Neural Network?



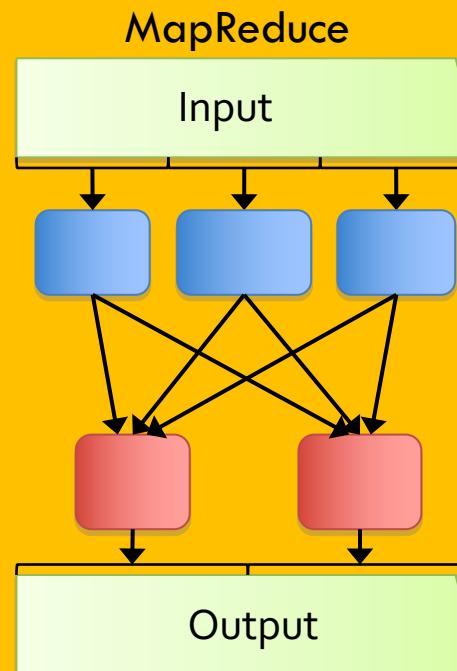
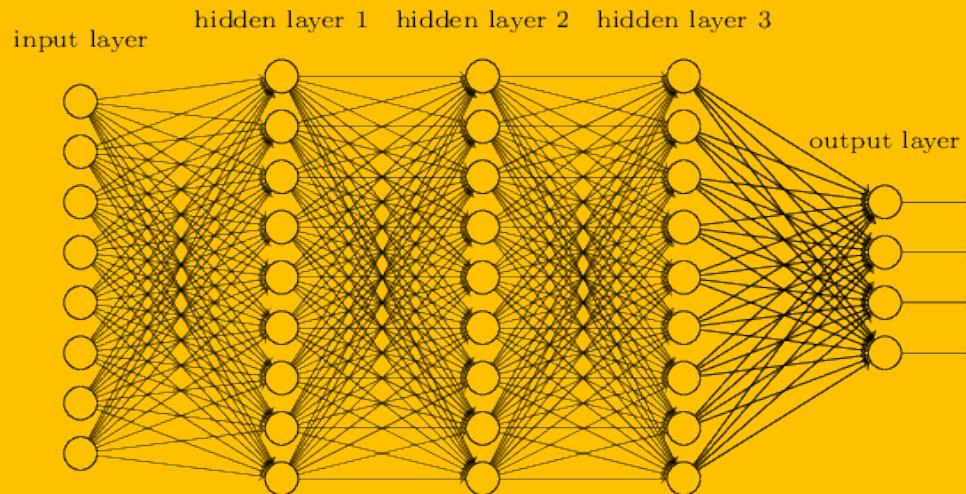
# Why a Parallel Neural Network?



Typical structure of an NN:

- For each training session
  - For each training example in the session
    - For each layer (forward and backward)
    - For each neuron in the layer
      - For all weights of the neuron
      - For all bits of the weight value

# Why a Parallel Neural Network?



# The Traveling ~~Salesman~~ Santa Problem



## Elastic Network

*is a type of neural network which utilizes unsupervised learning algorithms for clusterization problems and treats neural networks as a ring of nodes.*

*The learning process keeps changing the shape of the ring, where each shape represents a possible solution*

# The Traveling ~~Salesman~~ Santa Problem



Parallelized neural network system for solving Euclidean traveling salesman problem

Bihter Avşar<sup>a</sup>, Danial Esmaeili Aliabadi<sup>a,\*</sup>

<sup>a</sup>*Sabancı University, Faculty of Engineering and Natural Science, Istanbul, Turkey.*

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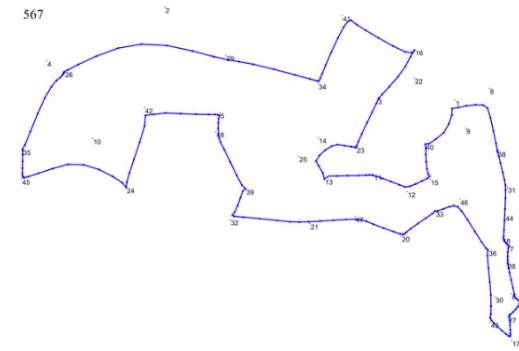
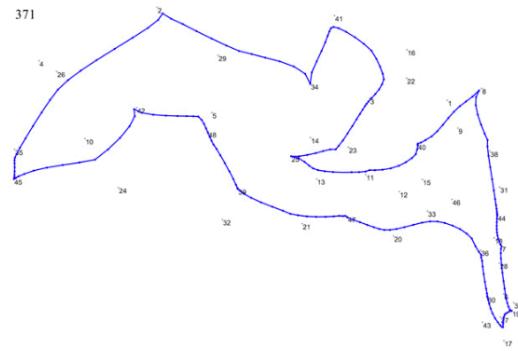
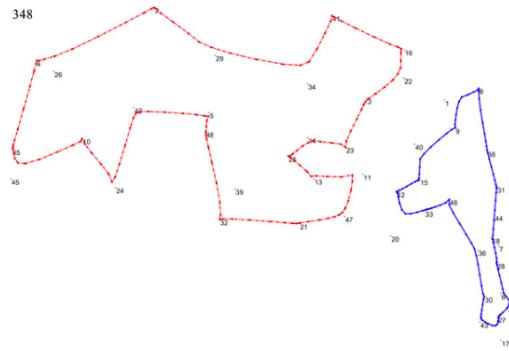
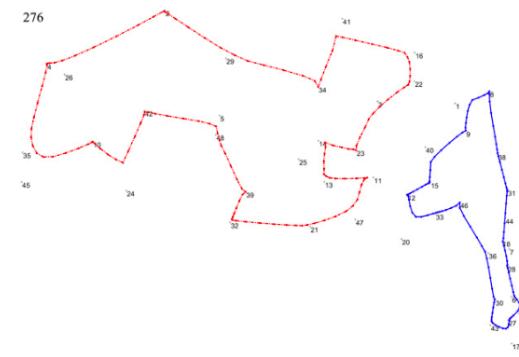
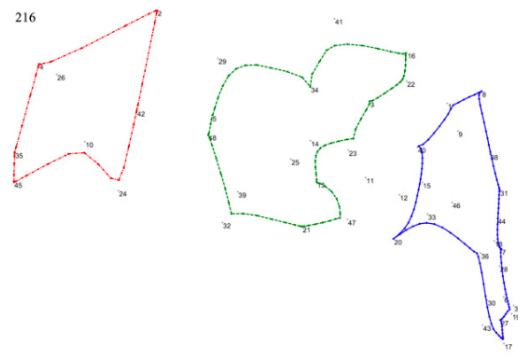
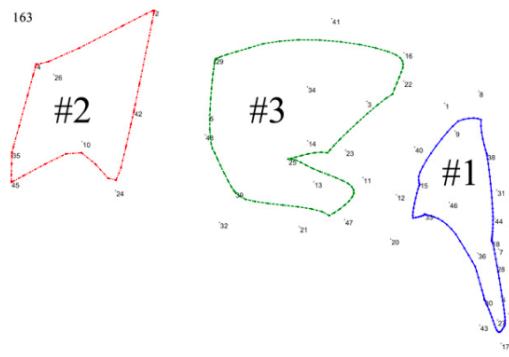
## Abstract

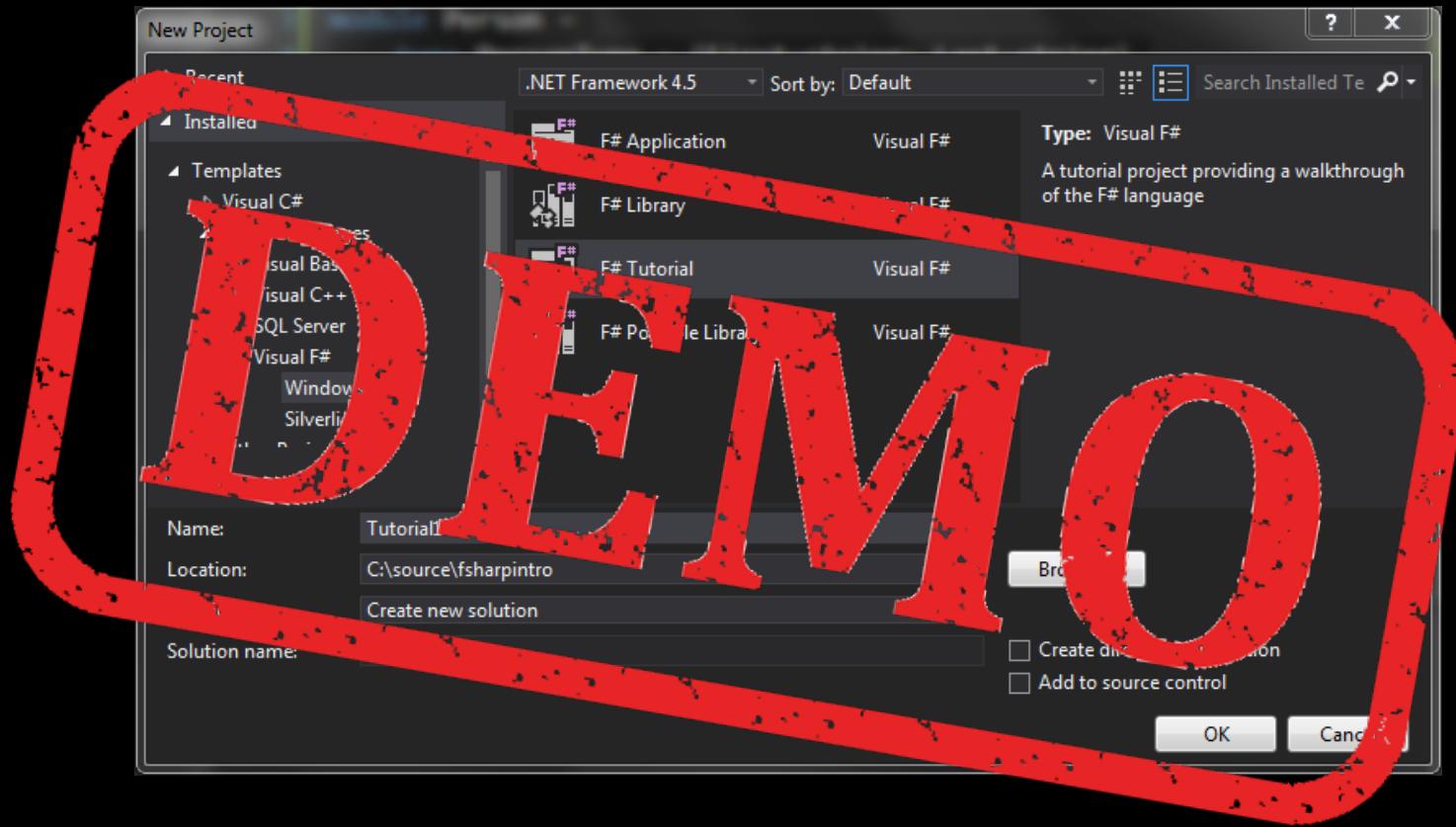
We investigate a parallelized divide-and-conquer approach based on a self-organizing map (SOM) in order to solve the Euclidean Traveling Salesman Problem (TSP). Our approach consists of dividing cities into municipalities, evolving the most appropriate solution from each municipality so as to find the best overall solution and, finally, joining neighborhood municipalities by using a blend operator to identify the final solution. We evaluate performance of parallelized approach over standard TSP test problems (TSPLIB) to show that our approach gives a better answer in terms of quality and time rather than the sequential evolutionary SOM.

*Keywords:* Euclidean Traveling Salesman Problem, Artificial Neural Network, Parallelization, Self-organized map, TSPLIB

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# The Traveling ~~Salesman~~ Santa Problem





# Summary



Neural Networks  
ease complex  
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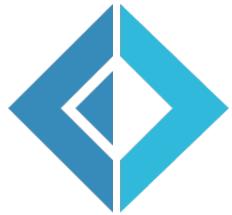
# Resources



- <ftp://ftp.sas.com/pub/neural/FAQ.html#questions>
- [Machine Learning Projects for .NET Developers](#)
- [Mastering .NET Machine Learning](#)
- [Evelina Gabasova blog](#)
- [James McCaffrey blog](#)



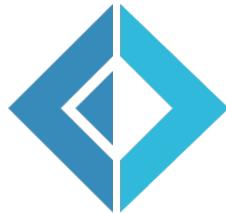
That's all Folks!



*The tools we use have a profound (and devious!) influence on our thinking habits, and, therefore, on our thinking abilities.*

-- Edsger Dijkstra

# How to reach me



<https://github.com/rikace/ml-workshop>

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