

Hacker's Guide to Neural Networks a. k. a. **Data Driven Code - 101**

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WHAT WE WILL **NOT** COVER

- Recent developments in
 - United States of America! ;)
 - Deep Learning, CNN, RNN, DCGAN etc.
 - Frameworks like TensorFlow, Theano, Keras etc.
- Advanced concepts of neural networks



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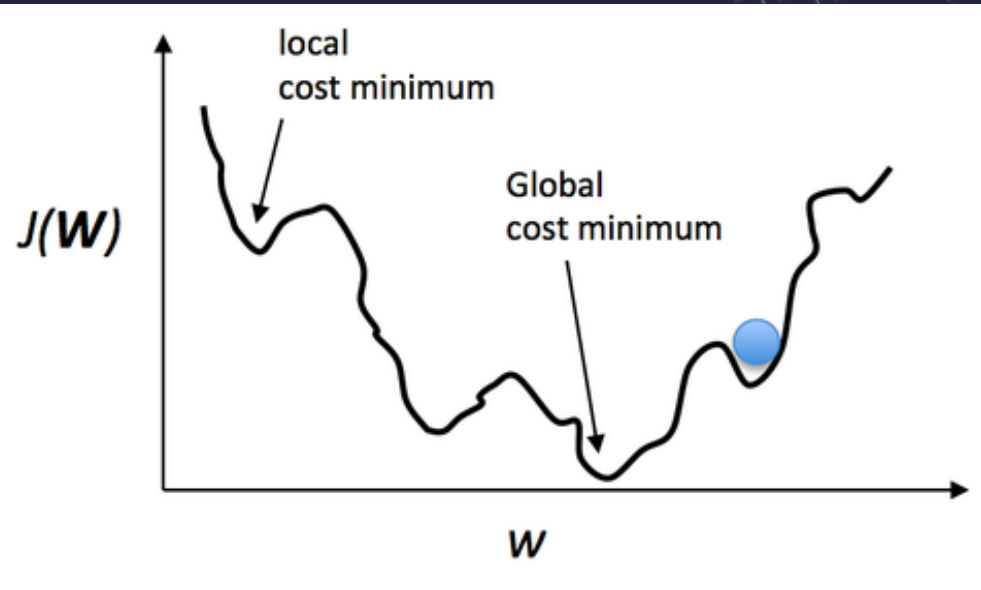
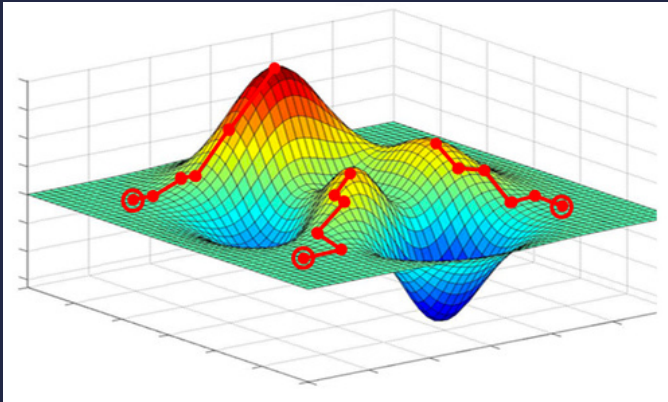
WHAT WE WILL COVER

- Basic Concepts of
 - Parameter Optimization
 - Entropy (Sparse Coding)
- Little bit of Mathematics
 - Linear Algebra – Matrix Multiplication
 - Differential Calculus – Sigmoid Function (2 equations only)
- (Try to) Build a 2 layer neural network
- Way Forward



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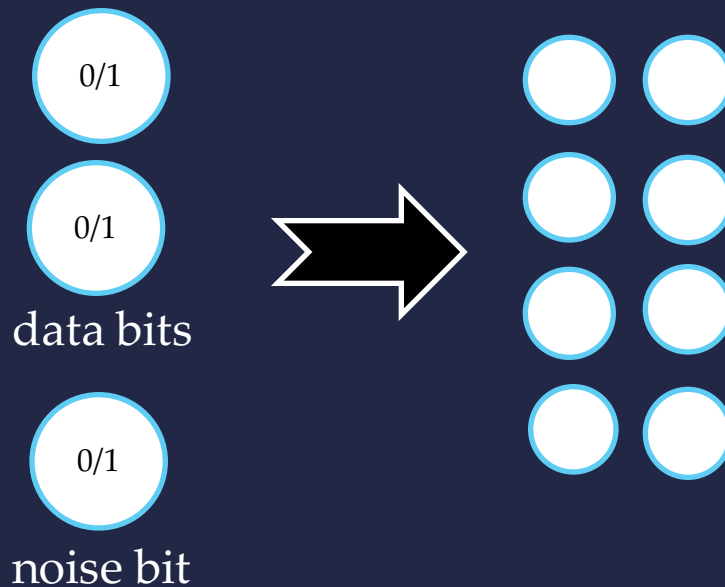
PARAMETER OPTIMIZATION



- Infinite Monkey Theorem
- Any problem is fundamentally a parameter optimization problem

ENTROPY

- Sparse Coding



$$2^8 = 128$$

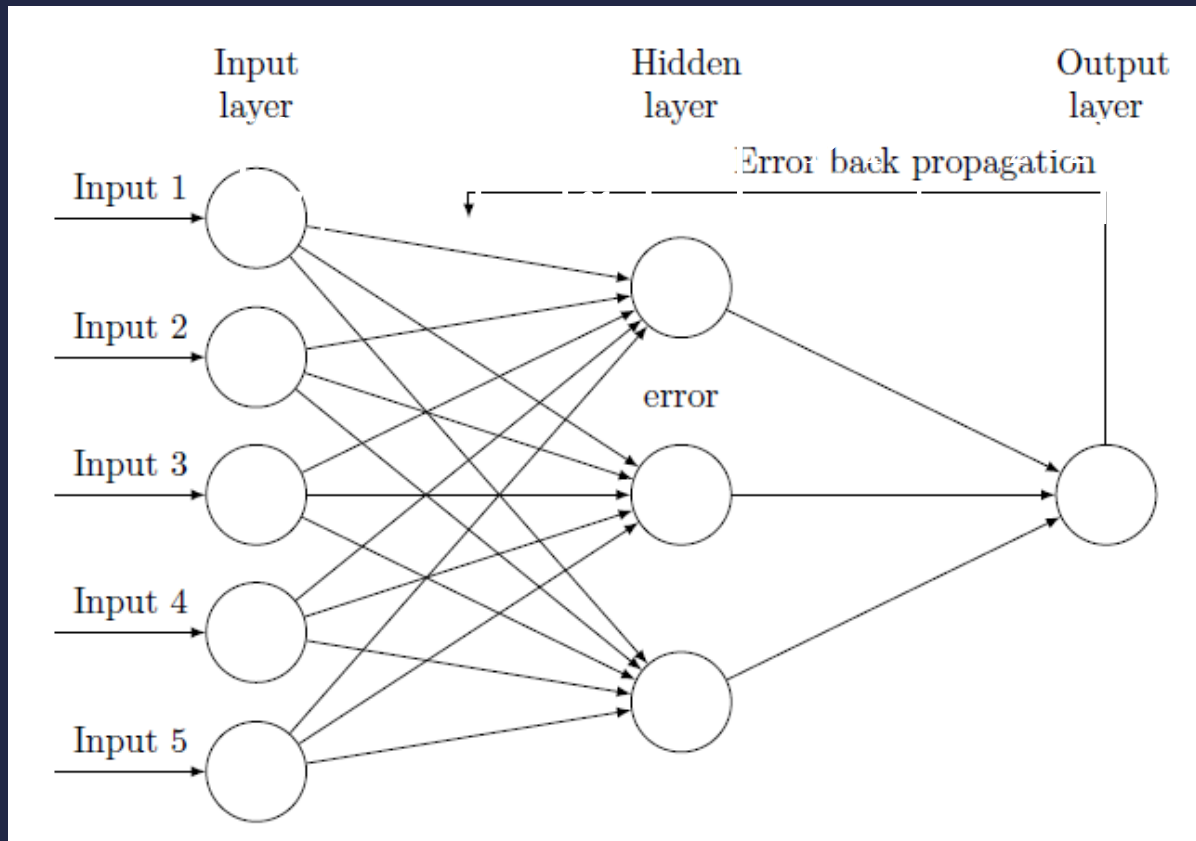
Imagine:

- divided into 4 spaces
- each bit – a feature
- generalize data bits
- attenuates noise bit



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LINEAR ALGEBRA PART



LINEAR ALGEBRA PART (matrix multiplication)

$$\begin{bmatrix} w_1 & w_2 \\ w_3 & w_4 \end{bmatrix} \times \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} w_1x_1 + w_2x_2 \\ w_3x_1 + w_4x_2 \end{bmatrix}$$



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ACTIVATION FUNCTION (differential calculus part)

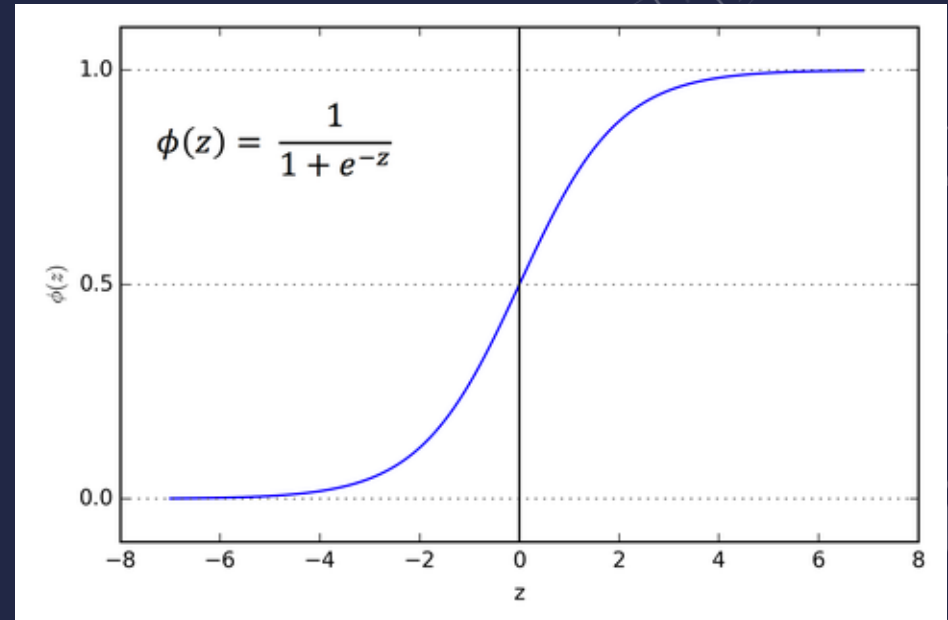
$$\sigma(x) = \frac{1}{1+e^{-x}}$$

(Sigmoid Function)

$$\frac{d\sigma(x)}{dx} = \sigma(x) \times (1-\sigma(x))$$

(Derivative of Sigmoid Function)

Stochastic Gradient Decent



QUICK RECAP

- Layers
- Dataset
- Training
- Network
- Features
- Sparse Encoding
- Back Propagation
- Activation Function
- Parameter Optimization

LET'S MAKE ONE NOW !

CODE + DATA = MAGIC

(Teach machine **XOR** truth table)

XOR GATE

Table 5.8 : Truth table for XOR Gate

INPUTS		OUTPUTS
A	B	$Y = A \oplus B$
0	0	0
0	1	1
1	0	1
1	1	0



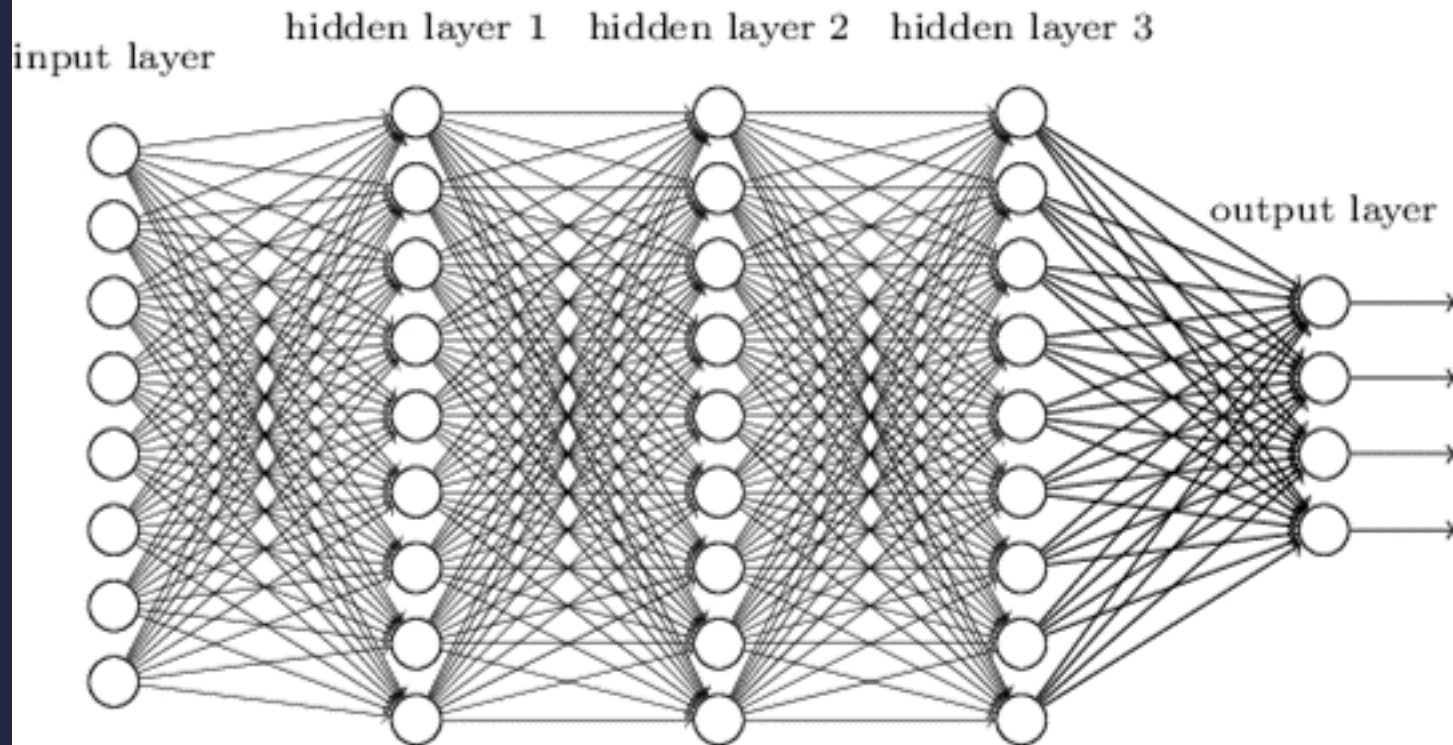
THE CODE

```
1  from numpy import array, random, dot, exp
2
3  inputs = array([[0, 0], [0, 1], [1, 0], [1, 1]])
4  outputs = array([[0], [1], [1], [0]])
5
6  def learn(X, y):
7      l1_w = 2 * random.random((X.shape[1], 16)) - 1
8      l2_w = 2 * random.random((16, 1)) - 1
9      for j in xrange(10000):
10         l1 = 1 / (1 + exp(-(dot(X, l1_w))))
11         l2 = 1 / (1 + exp(-(dot(l1, l2_w))))
12         l2_delta = (y - l2) * (l2 * (1 - l2))
13         l1_delta = l2_delta.dot(l2_w.T) * (l1 * (1-l1))
14         l2_w += l1.T.dot(l2_delta)
15         l1_w += X.T.dot(l1_delta)
16     return (l1_w, l2_w)
17
18  xor_weights = learn(inputs, outputs)
19
20  def predict(X, weights):
21      l1 = 1/(1+exp(-(dot(X, weights[0]))))
22      l2 = 1/(1+exp(-(dot(l1, weights[1]))))
23      return l2
24
25  test_set = [[0, 0], [0, 1], [1, 0], [1, 1]]
26  for test_item in test_set:
27      xor_prediction = predict(test_item, xor_weights)
28      print str(test_item)+"\t"+str(xor_prediction)
```



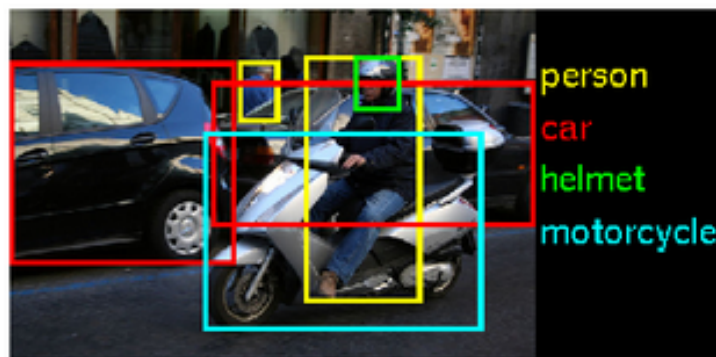
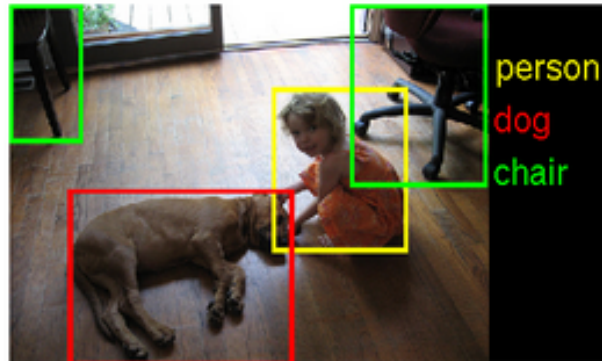
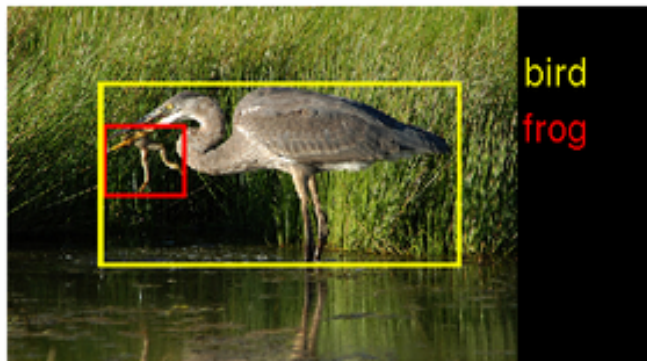
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Deep neural network

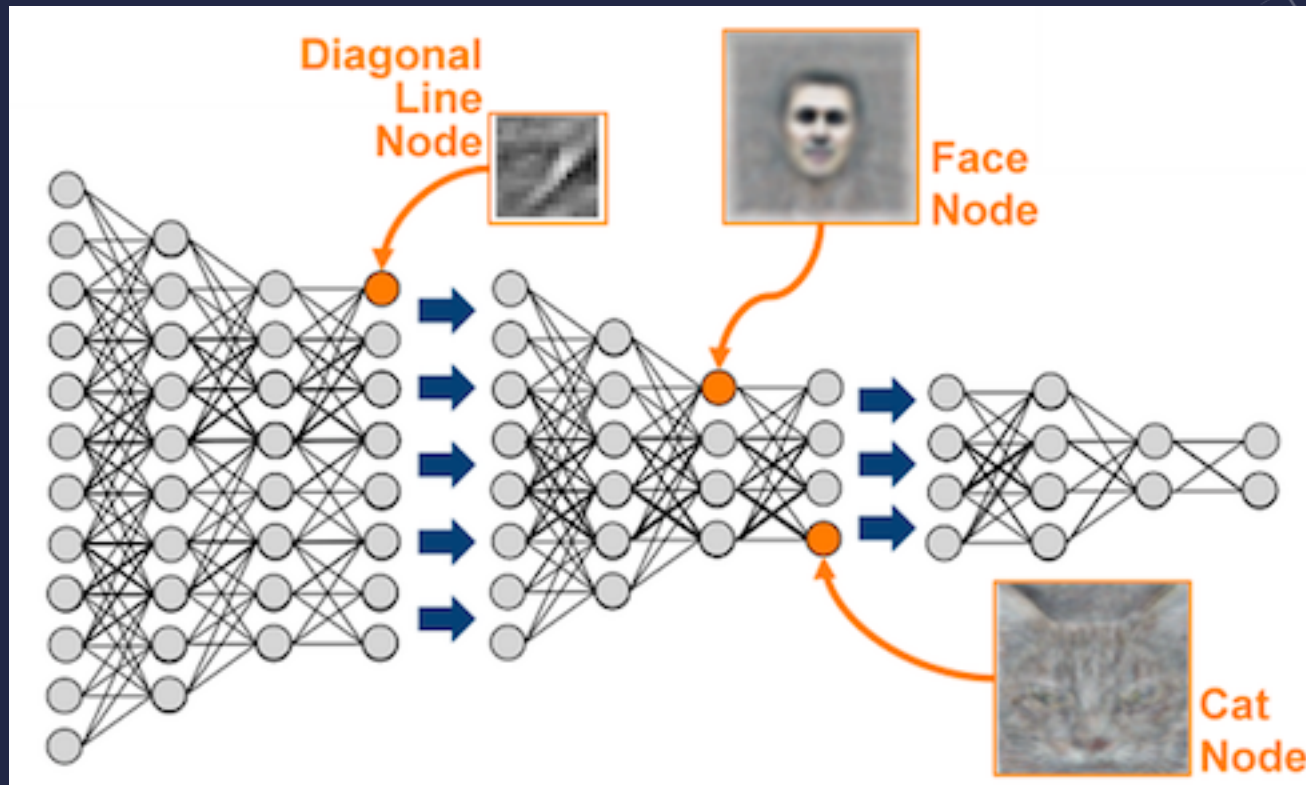


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DEEP LEARNING



DEEP LEARNING



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WAY FORWARD

Play with Neural networks in the browser

<http://playground.tensorflow.org/>

Comprehensive list of resources available online on Deep Learning

<https://github.com/ChristosChristofidis/awesome-deep-learning>

REFERENCE LINKS / IMAGE COURTESY

- <https://medium.com/technology-invention-and-more/how-to-build-a-simple-neural-network-in-9-lines-of-python-code-cc8f23647ca1>
- <https://medium.com/technology-invention-and-more/how-to-build-a-multi-layered-neural-network-in-python-53ec3d1d326a#.dis92by0e>
- <http://iamtrask.github.io/2015/07/12/basic-python-network/>
- <http://iamtrask.github.io/2015/07/27/python-network-part2/>
- <http://sebastianraschka.com/faq/docs/logisticregr-neuralnet.html>



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THANK YOU

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- Code - <https://github.com/atmb4u/data-driven-code>
- Slides – <https://slideshare.net/atmb4u/data-driven-code>



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