

# Applied Data science

Spring 2017\_Project 3\_Group 14

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# Models Comparison

feature	model	parameters	accuracy	Time
sift	GBM	ntree = 100, depth =1, shrinkage = 0.1 (CV)	0.73	21.5s
sift (LASSO)	GBM	ntree = 100, depth =1, shrinkage = 0.1 (CV)	0.64	20s
sift (LASSO)	KNN	k = 3	0.65	10min
sift (LASSO)	XG boost	objective = 'logistic', max_depth = 7, eta = 0.11, gamma = 0.01	0.68	40s
sift (LASSO)	random forest	ntree = 600 (CV)	0.67	5min
sift (LASSO) + texture	majority vote (XG boost)	...	0.69	18min
DNN	DNN	num.round = 750, learning.rate = 0.3, dropout = 0.7	0.81	15min (features) + 5min(model)

Training: 80% vs. Test: 81.24% improved nearly 9%

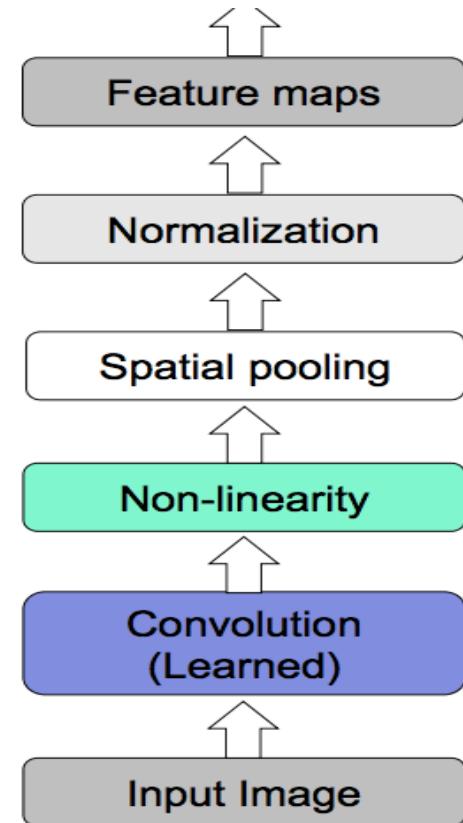


# Deep Learning (DNN)

**Deep learning: “Deep” architecture**



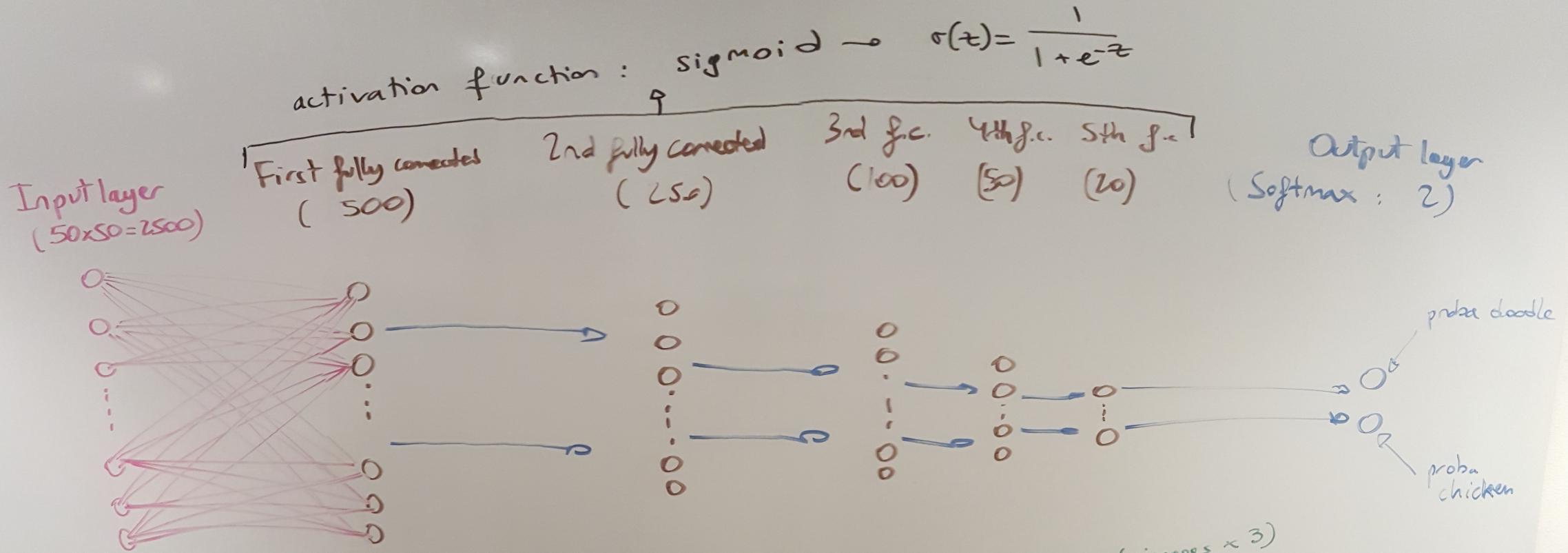
# Feature Extractions



# Problems + Strategies

- 2000 images -> 1600 train + 400 test
- Deep Learning (mxnet)
- Results: train acc=95% vs. test acc=75%
- Problem: Overfitting !!!
- Reason: Deep networks need to be trained on a huge number of training images to achieve satisfactory performance
- How to solve? Ans: data augmentation (add to the general training dataset images that have been flipped horizontally and also with one rotation of small angles)
- Strategies: 2000 images -> 4000 + 2000 images, improved!





### Most important points ::

- Total numbers of parameters::  
 matrices {  $2500 \times 500 + 500 \times 250 + 250 \times 100$   
 $+ 100 \times 50 + 50 \times 20 + 20 \times 2$   
 bias {  $1 + 500 + 250 + 100 + 50 + 20$

$\approx 1.4 \text{ M}$  parameters,

- Training accuracy  $\approx 80\%$ .
- Test accuracy  $\approx 79.24\%$ .

$\rightarrow$  time to train  $\approx 20$  minutes

on CPU

of  
MacBook Pro 2013

① Data augmentation to avoid overfitting (images  $\times 3$ )

② Dropout to avoid massively overfitting (set to  $p=0.9$ )

③ Larger learning rate to avoid getting stuck in a "hole"  $\nabla J$

④ Larger weights initialization matrix  $W$ :  $W_{ij} \sim N(0, 0.6)$

↳ original initialization diverges for  $W_{ij} \approx 0$

⑤ Use of sigmoid activation function

↳ tanh and relu all diverge

# Final DNN model

```
##  
mx.set.seed(2)  
model <- mx.mlp(train.x, train.y, hidden_node=c(500,250,100,50,20), out_node=2, activation =  
"sigmoid", out_activation="softmax",  
    num.round=200, array.batch.size=50, learning.rate=0.08,  
    eval.metric=mx.metric.accuracy, dropout = 0.9, momentum = 0.7,  
initializer=mx.init.normal(0.6))  
  
accuracy_model(model, test.x,test.y) # test accuracy 81.24%
```

```
[199] Train-accuracy=0.8172910000000001  
[200] Train-accuracy=0.834583333333334  
Auto detect layout of input matrix, use rowmajor..  
      test.y  
pred.label  0   1  
      0 445 109  
      1 128 518  
[1] "Accuracy of model: 80.25%"
```

