

# Flavor tagging based on deep learning

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01

The relationship among AI, machine learning  
and deep learning

# AI, Machine Learning AND Deep Learning



## AI (Artificial Intelligence)

AI is intelligence exhibited by machines.



## Machine Learning

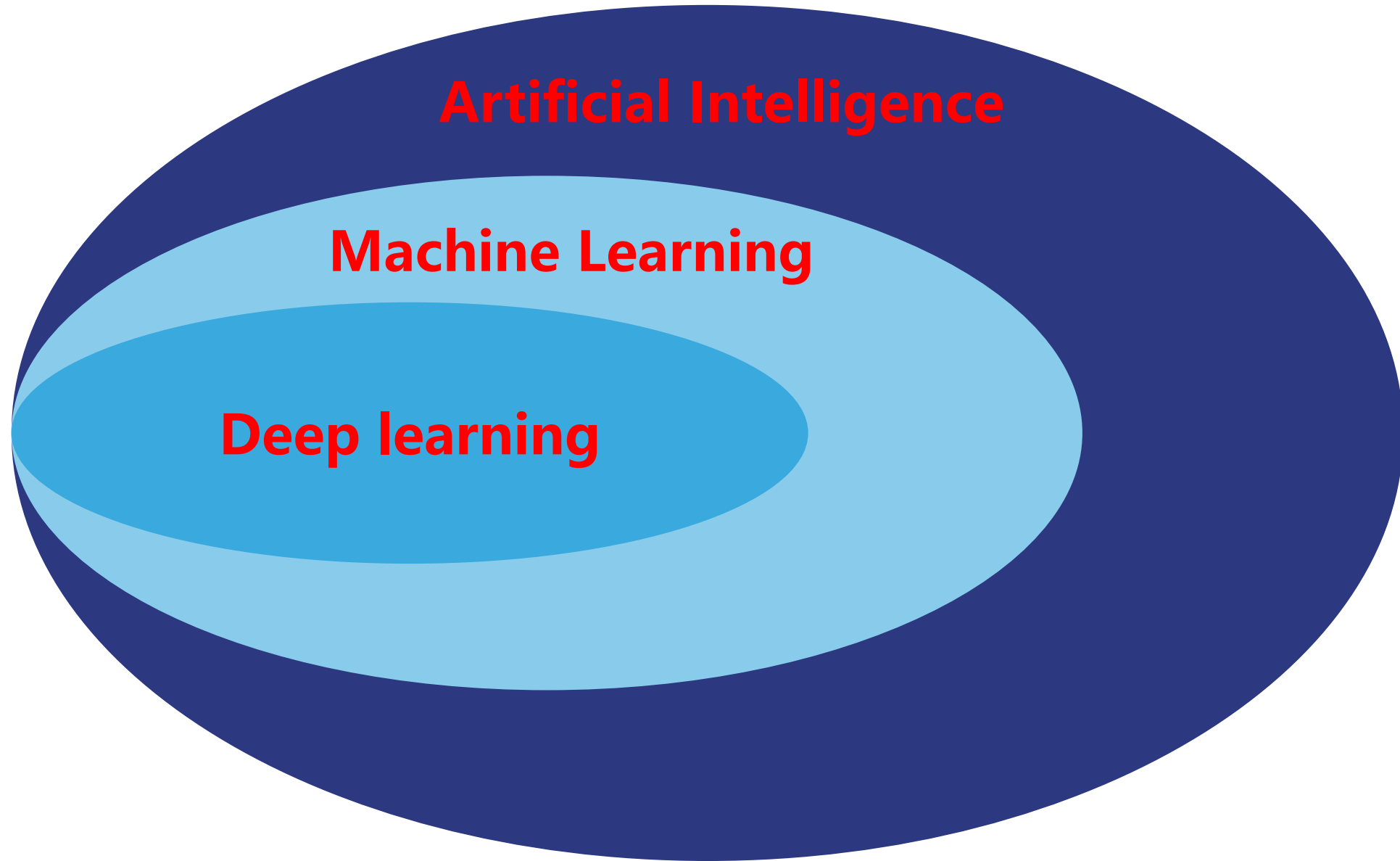
Machine learning is a model to realize AI.



## Deep Learning

DL is an important way of machine learning.

# AI, Machine Learning AND Deep Learning

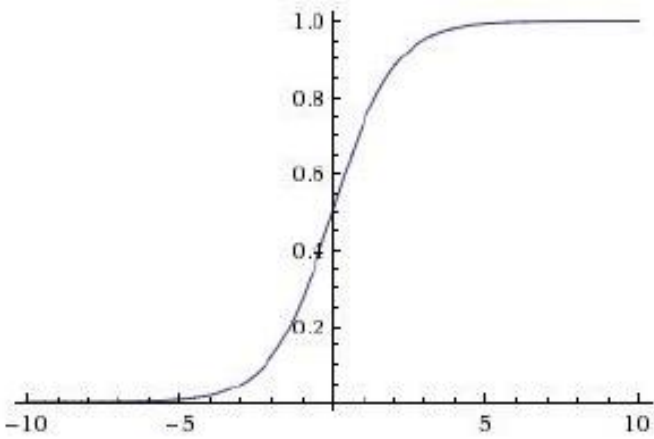
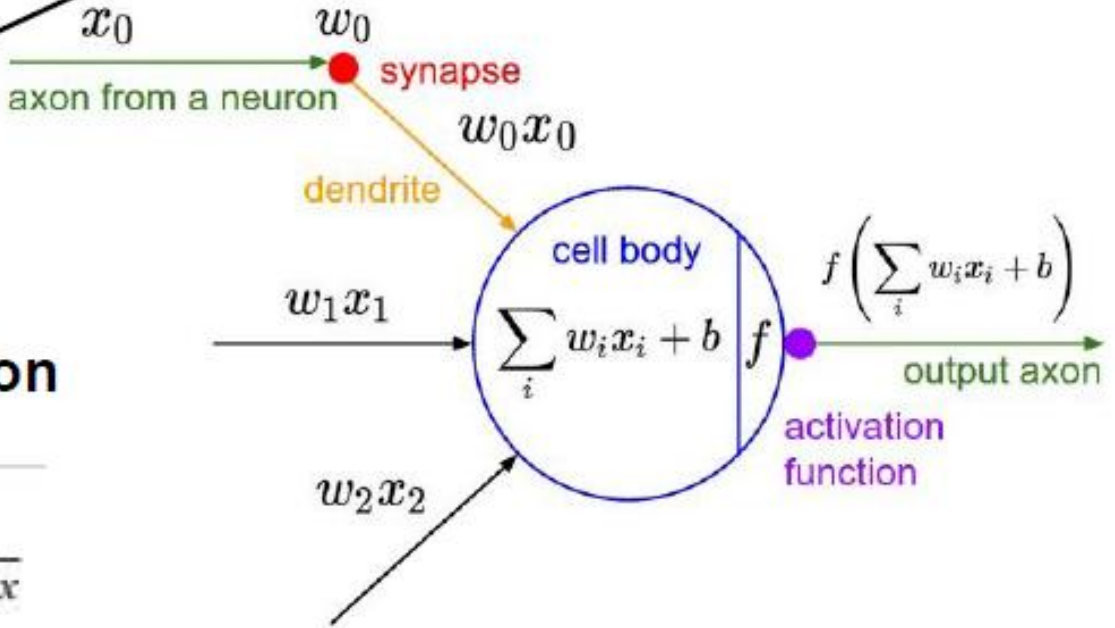
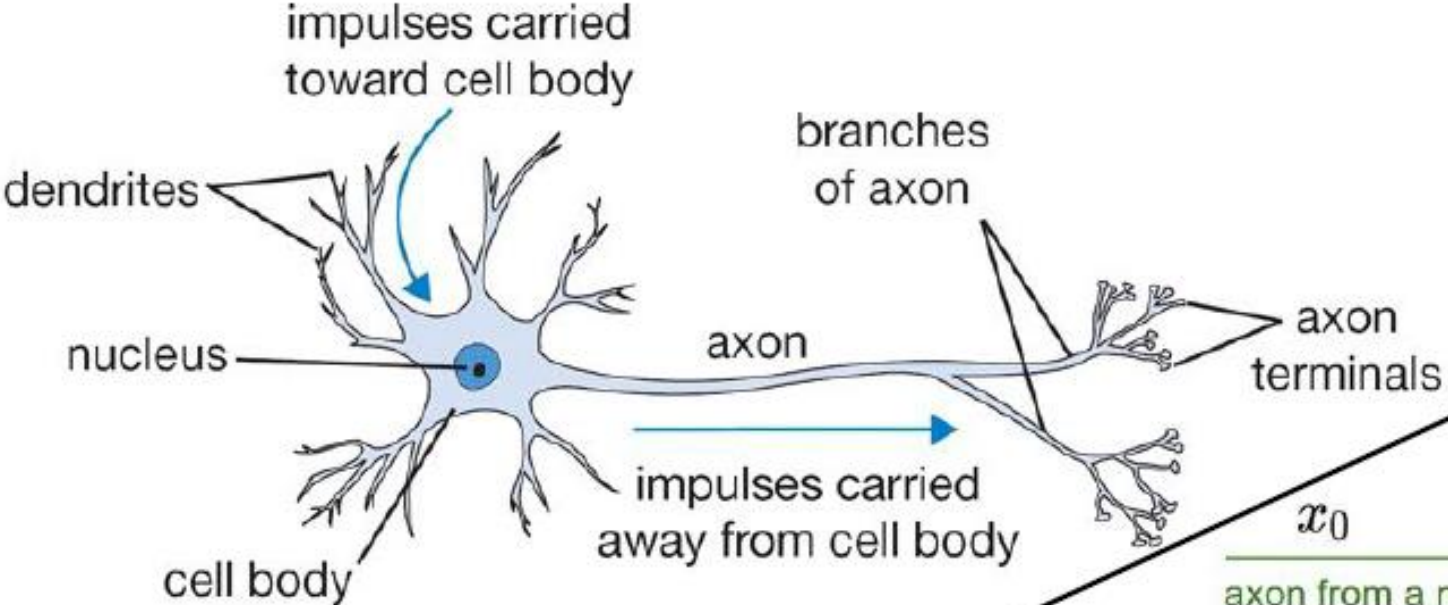
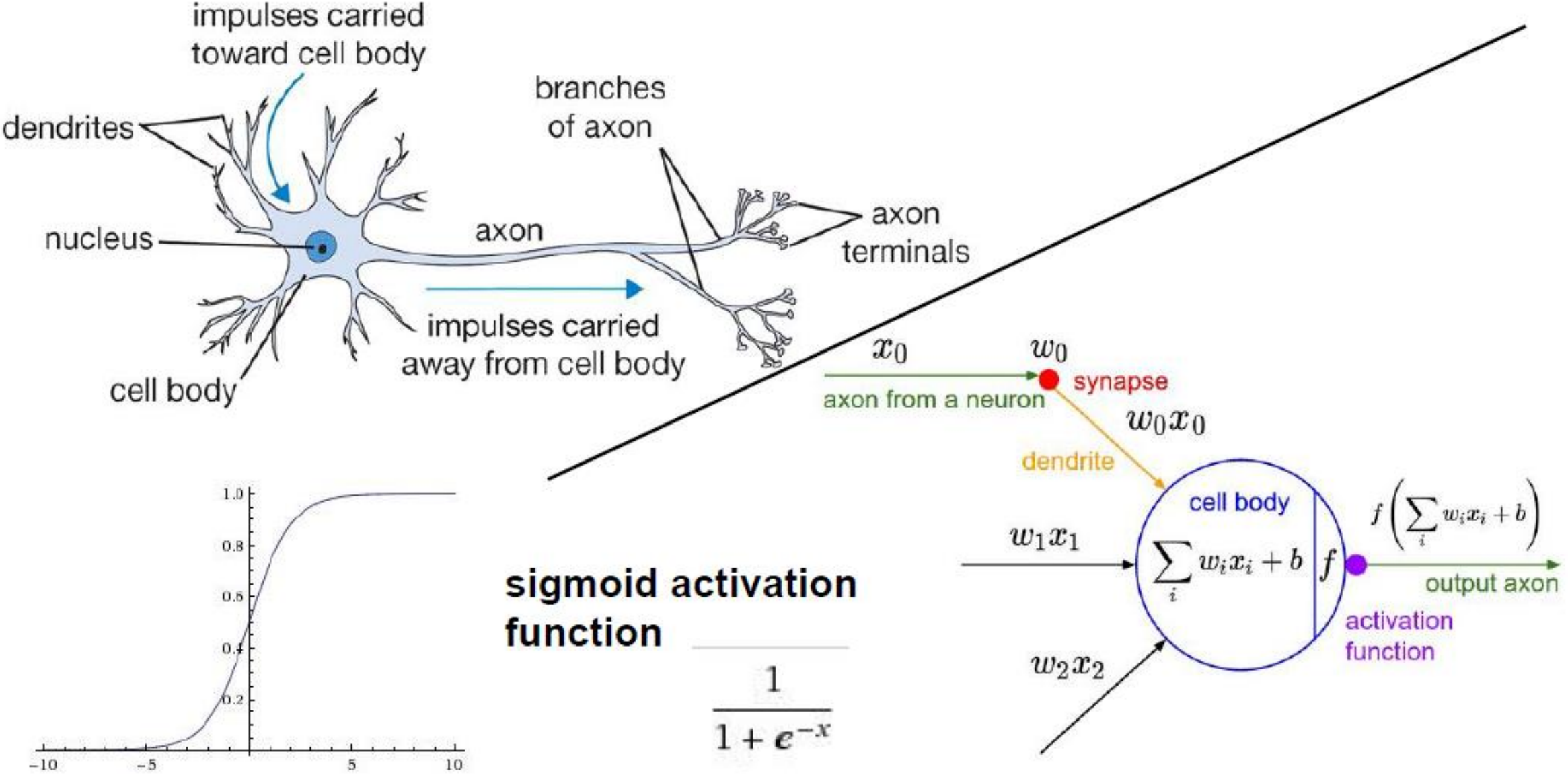




02

A brief introduction of deep learning

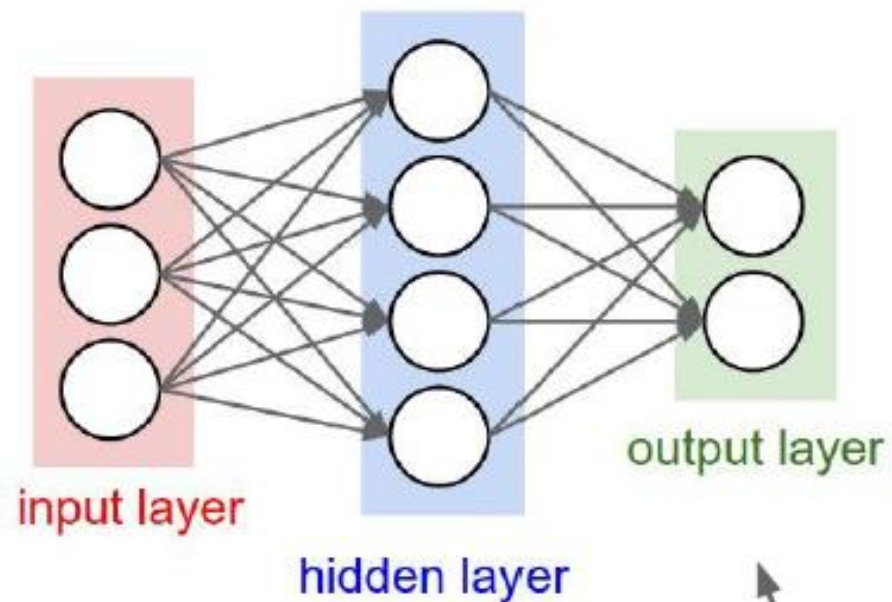
# The basic element of neural network



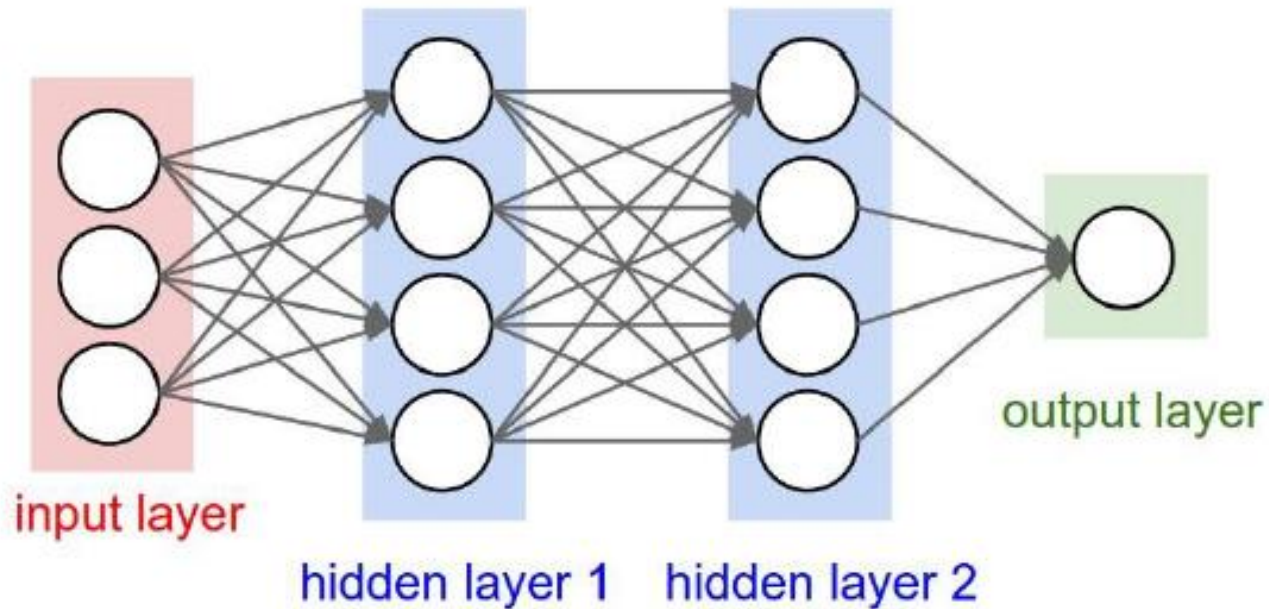
**sigmoid activation function**

$$\frac{1}{1 + e^{-x}}$$

# Architectures of neural networks



"2-layer Neural Net", or  
"1-hidden-layer Neural Net"

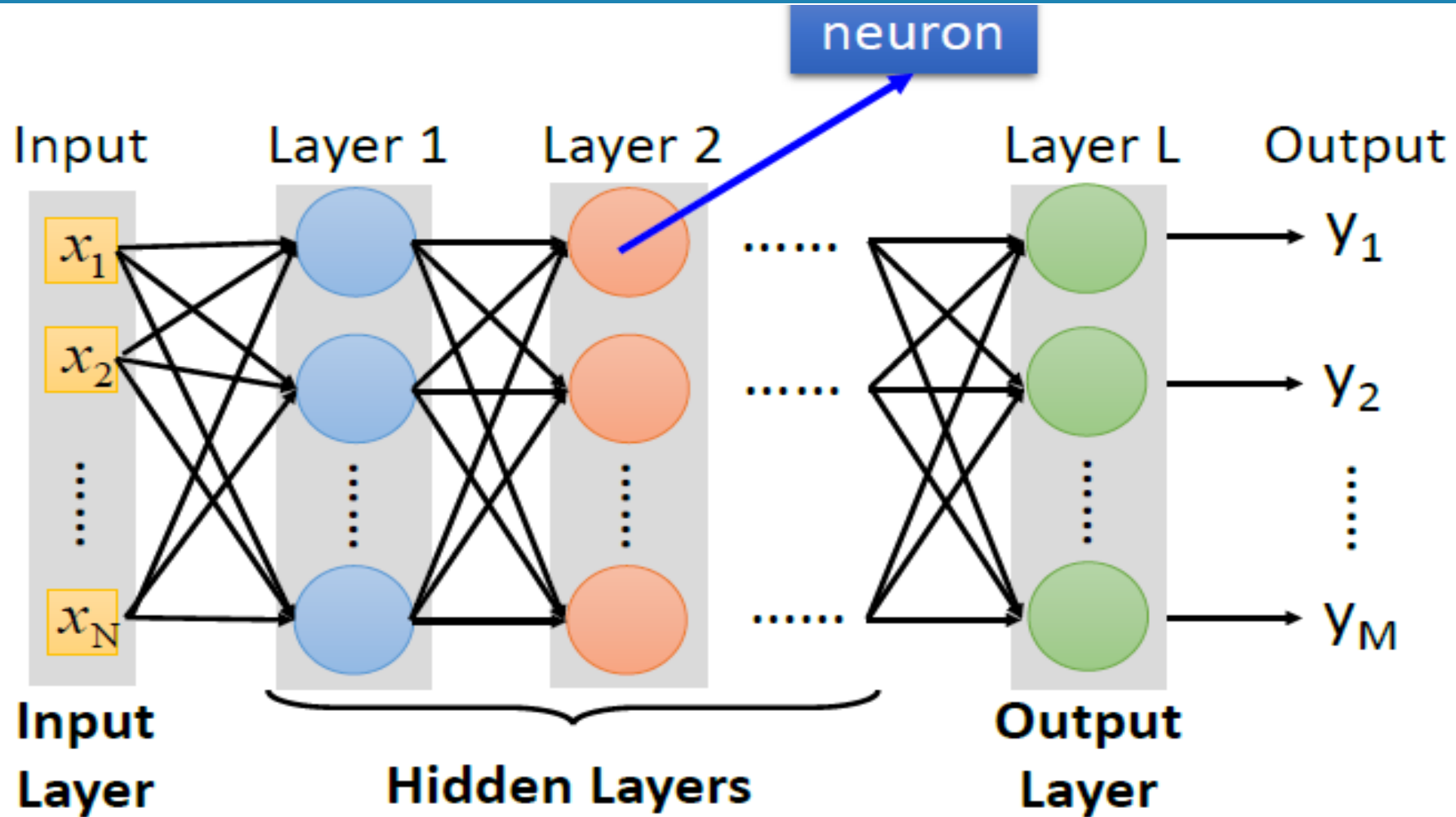


"3-layer Neural Net", or  
"2-hidden-layer Neural Net"

**"Fully-connected" layers**



# Deep neural network(DNN):Many hidden layers.



Deep means many hidden layers

03

Apply DNN to the flavor tag work

# Flavor tag(VTX\_08mm):

**01**

## Data set

200k training events

50k test events

**03**

## Output

Three outputs represent  
b,c and other tags.

**02**

## Input parameters

We totally have 68 parameters.

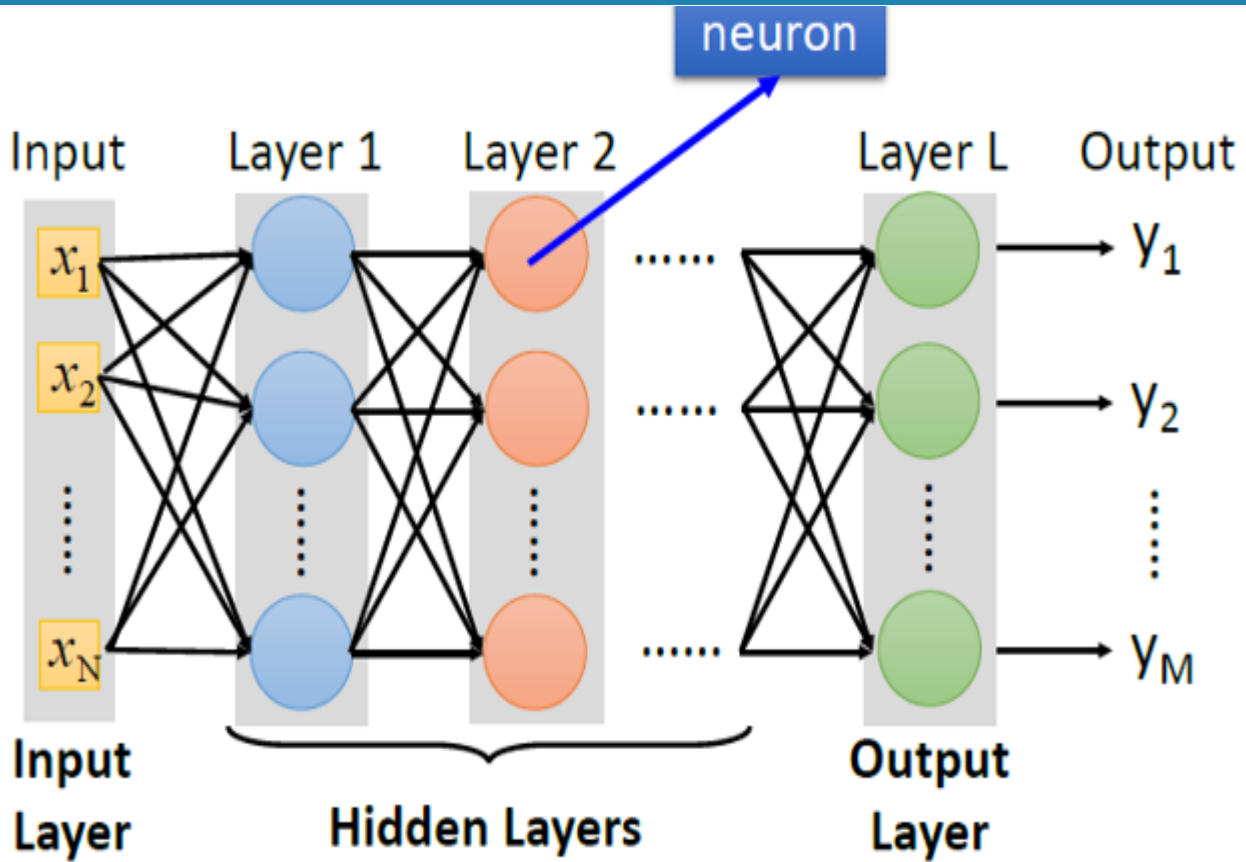
'aux','vtxlen1','vtxlen1\_j','z0cprob','d0  
bprob' are removed.

And then we have 63 inputs.

## Total parameters

```
['aux' 'ntrkwitho' 'ntrk' 'nvtxall' 'vtxmassal' 'vtxlen12a' 'vtxlen12a'  
'lvtxprob' 'nvtx' 'jete' 'jetcosine' 'jeteta' 'vtxlen1' 'vtxlen2'  
'vtxlen12' 'vtxlen1_j' 'vtxlen2_j' 'vtxlen12_' 'vtxsig1' 'vtxsig2'  
'vtxsig12' 'vtxsig1_j' 'vtxsig2_j' 'vtxsig12_' 'vtxdirang' 'vtxdirang'  
'vtxdirang' 'vtxdirang' 'vtxdirang' 'vtxdirang' 'vtxmom' 'vtxmom1'  
'vtxmom2' 'vtxmom_je' 'vtxmom1_j' 'vtxmom2_j' 'vtxmass' 'vtxmass1'  
'vtxmass2' 'vtxmasspc' 'vtxmult' 'vtxmult1' 'vtxmult2' 'vtxprob'  
'trk1d0sig' 'trk2d0sig' 'trk1z0sig' 'trk2z0sig' 'trk1pt' 'trk2pt'  
'trk1pt_je' 'trk2pt_je' 'jprobr' 'jprobz' 'jprobr5si' 'jprobz5si'  
'sphericit' 'Fd' 'jetrho' 'trkmass' 'nmuon' 'nelectron' 'd0bprob'  
'd0cprob' 'd0qprob' 'z0bprob' 'z0cprob' 'z0qprob']
```

# DNN structure for flavor tag



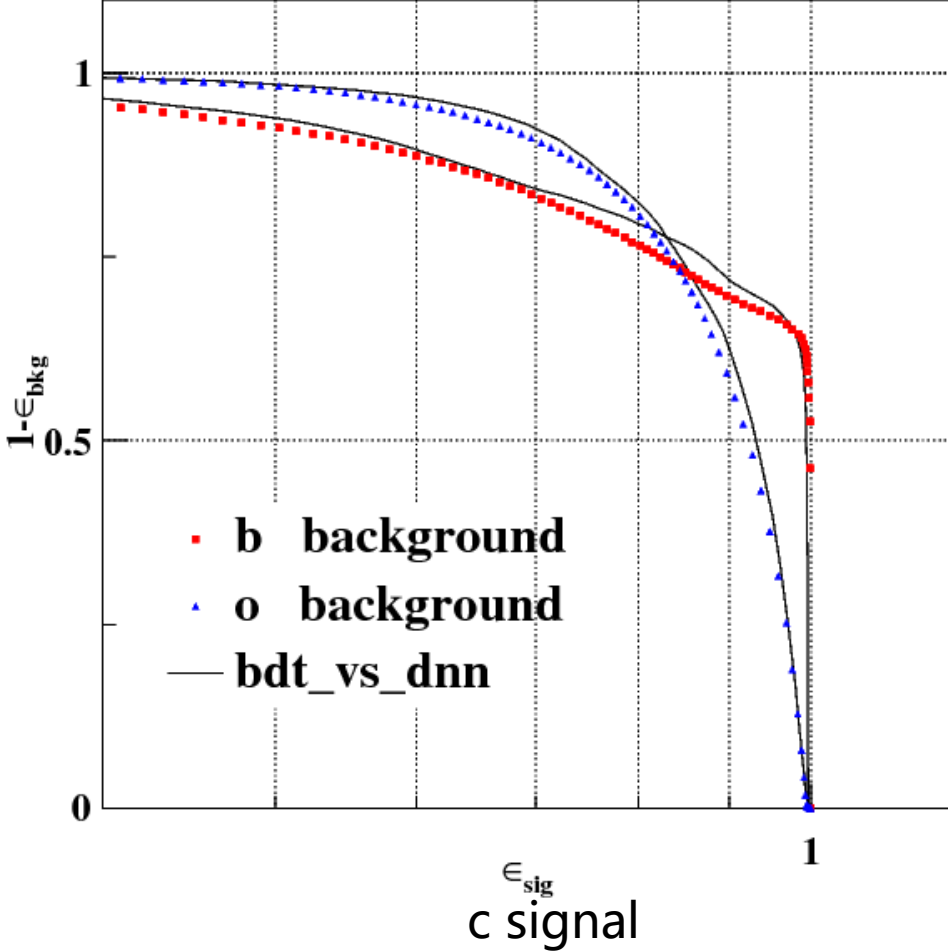
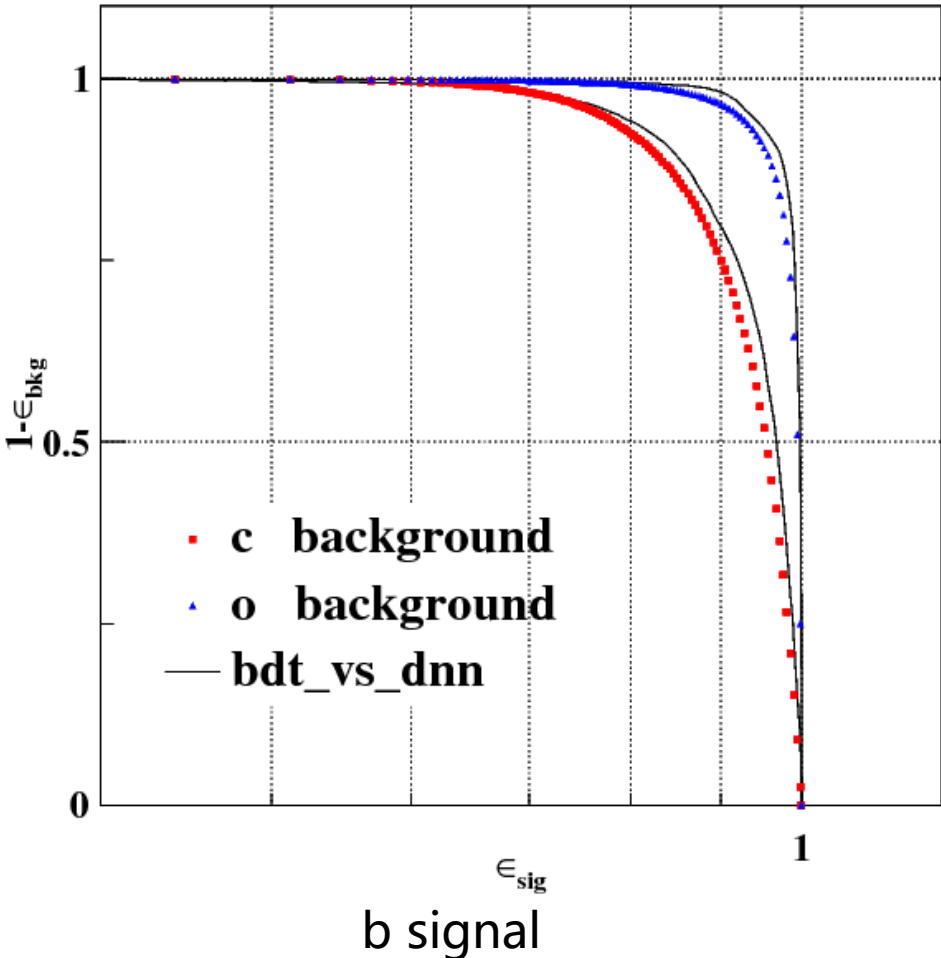
Layer 1 (Input layer): 63 neurons  
Layer 2: 512 neurons dropout: 0.5  
Layer 3: 512 neurons dropout: 0.5  
Layer 4: 512 neurons dropout: 0.5  
Layer 5: 512 neurons dropout: 0.5  
Layer 6: 256 neurons dropout: 0.5  
Layer 7: 256 neurons dropout: 0.5  
Output layer: 3 neurons

dropout=0.5 means that the probability that neurons are removed during training is 0.5

- The entire dataset of 250000 was normalized prior to any training.
- Network parameters were trained using gradient descent with mini-batches of 100 examples per step and 300000 steps.
- The neural networks were trained using the tensorflow package.

# ROC compared with bdt:

BDT:630k events DNN:250k events



# Signal Efficiency & Background Rejection

## b signal

$E_{sig}(b)$	$1-E_{bkg}(c, dnn)$	$1-E_{bkg}(c, bdt)$	$1-E_{bkg}(o, dnn)$	$1-E_{bkg}(o, bdt)$
0.8	0.94	0.92-0.93	0.99	0.99
0.9	0.79-0.80	0.75-0.76	0.98	0.96-0.97
0.95	0.62-0.65	0.52-0.55	0.93-0.94	0.90-0.91

## c signal

$E_{sig}(c)$	$1-E_{bkg}(b, dnn)$	$1-E_{bkg}(b, bdt)$	$1-E_{bkg}(o, dnn)$	$1-E_{bkg}(o, bdt)$
0.8	0.8	0.77	0.83	0.81
0.9	0.72	0.7	0.62	0.59
0.95	0.68	0.67	0.4	0.38

## Conclusion

For flavor tagging, DNN can improve the background rejection compared with BDT in the same signal efficiency.

Deep learning may be a promising method to deal with other classification problems with lots of physical parameters.