What is Deep Learning?

BSDS 100, Spring 2021 Michael Ruddy

What is Deep Learning?

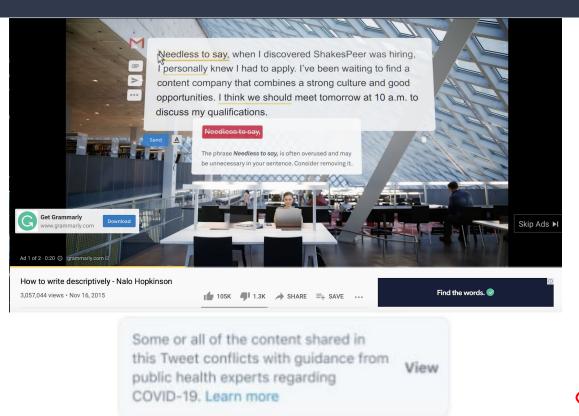
- Deep Learning often refers to a subset of Machine Learning algorithms that work by extracting *high-level features* from datasets that are often gigantic and/or largely unstructured
- Neural Networks are the workhorse behind many Deep Learning algorithms
 - Computer Vision
 - Natural Language Processing

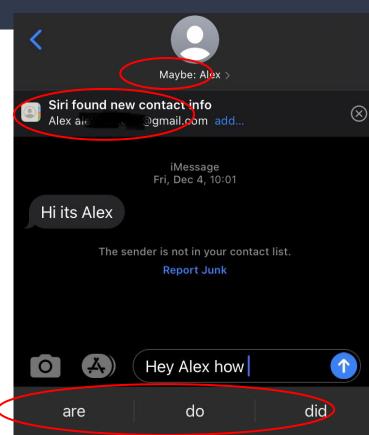
Result of partial training Image generation by DeepMind AI (a) 128×128 (b) 256×256 (c) 512×512 (d)

Large Scale GAN Training for High Fidelity Natural Image Synthesis, Brock A., Donahue J., Simonyan S., 2019



Tom Cruise Deepfake by@deeptomcruise on Tik Tok



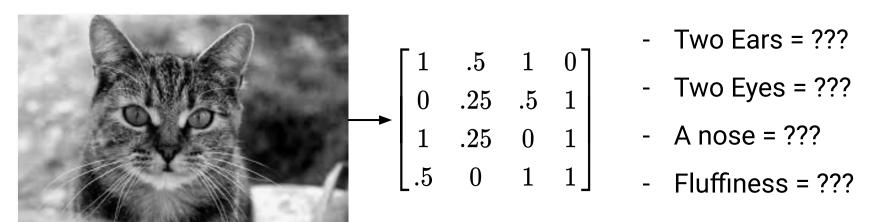


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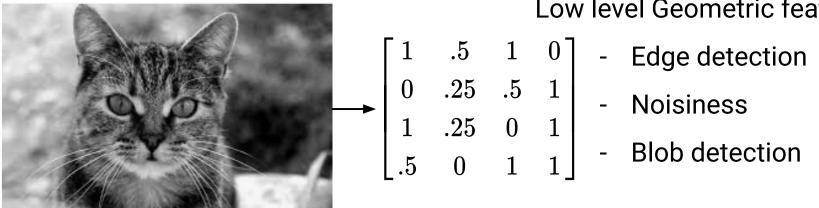
- Two Ears
- Two Eyes
- A nose
- Fluffiness ~.8

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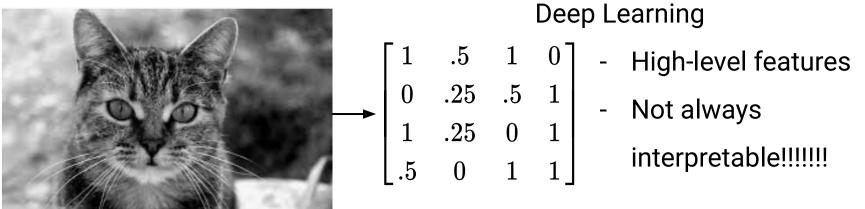
- Two Ears = ???

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Low level Geometric features

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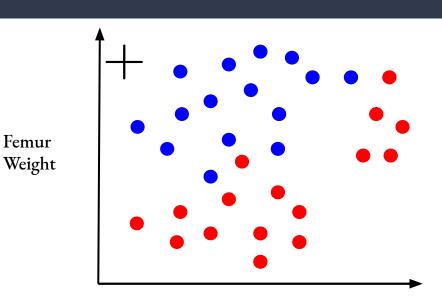


Deep Learning

- Task: Classify as Cat or Dog
- Data:
 - Femur Weight
 - Femur Length



Sabre Tooth Tiger Femurs, U of FL



Femur Length

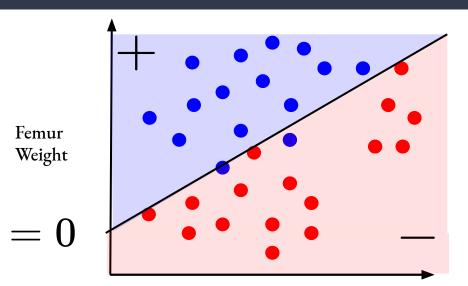


$$f=a_1+a_2(length)+a_3(weight)$$

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Femur Length

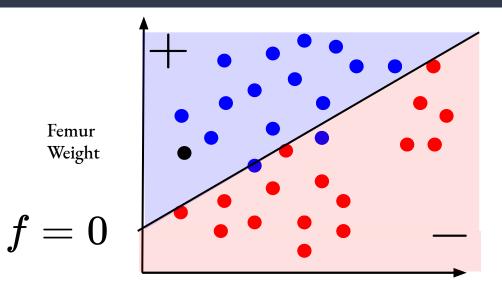


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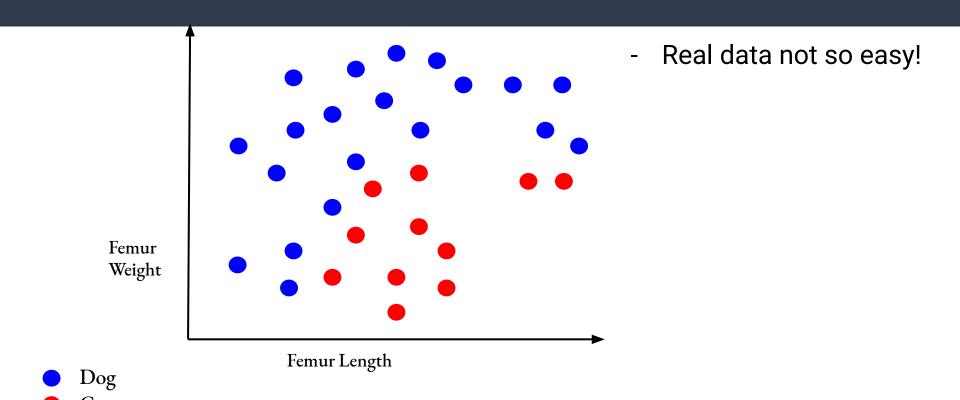


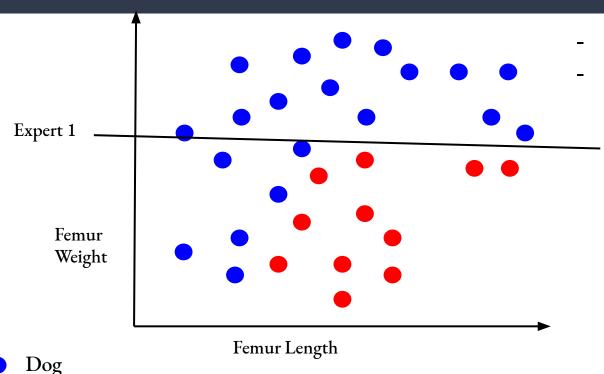
Sabre Tooth Tiger Femurs, U of FL



Femur Length

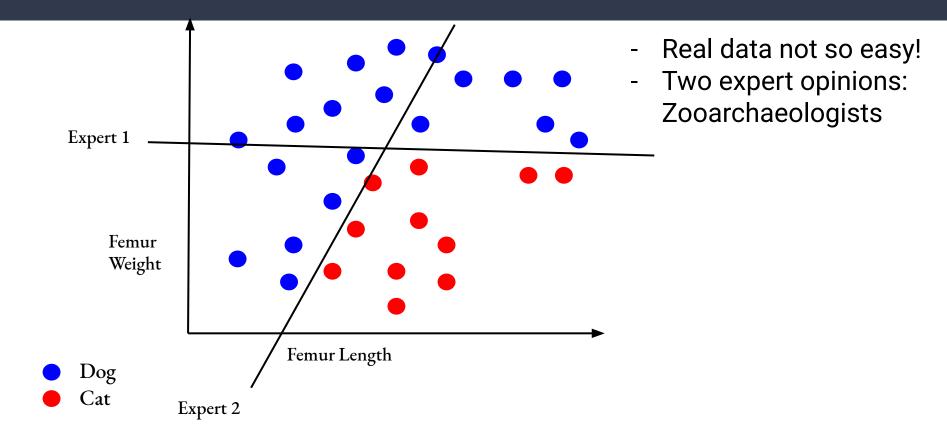


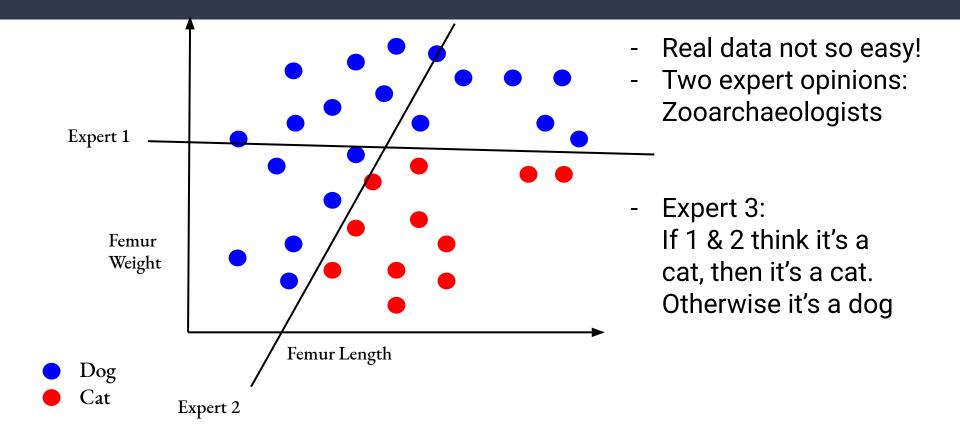


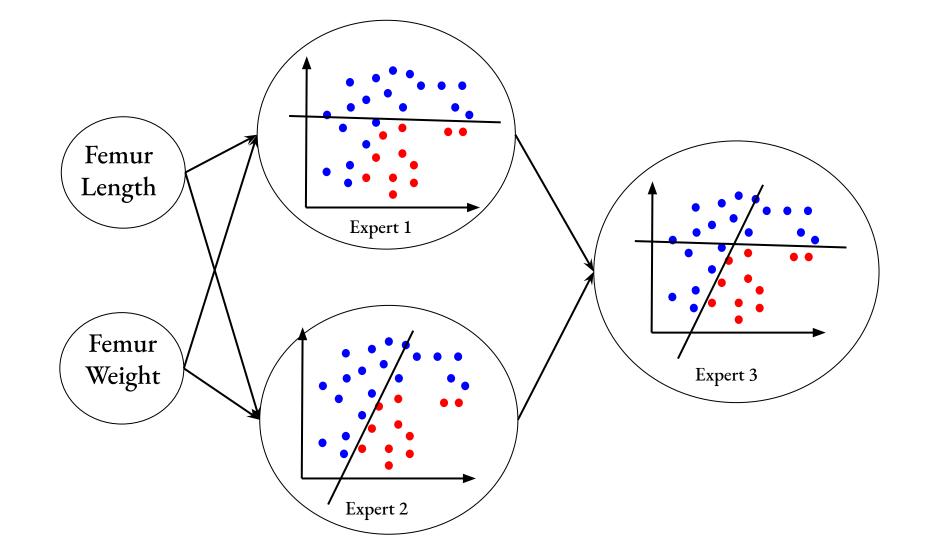


- Real data not so easy!

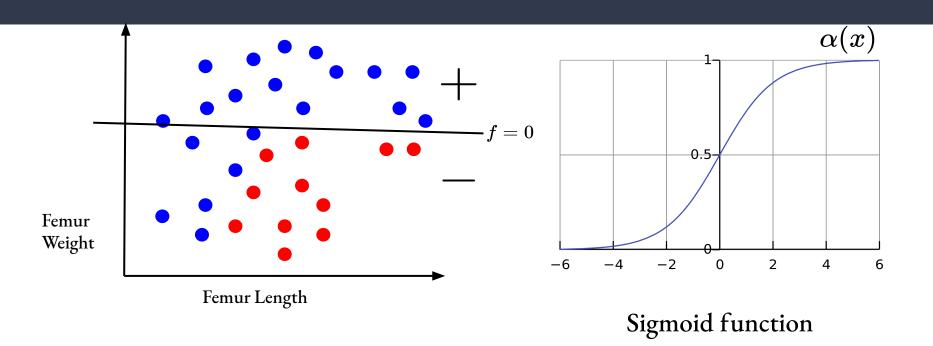
Two expert opinions:Zooarchaeologists





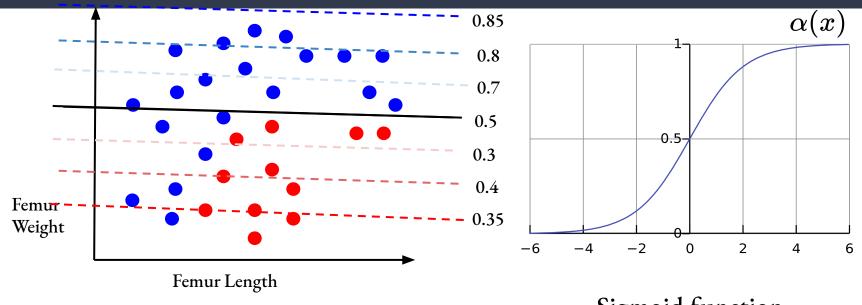


Dog Cat

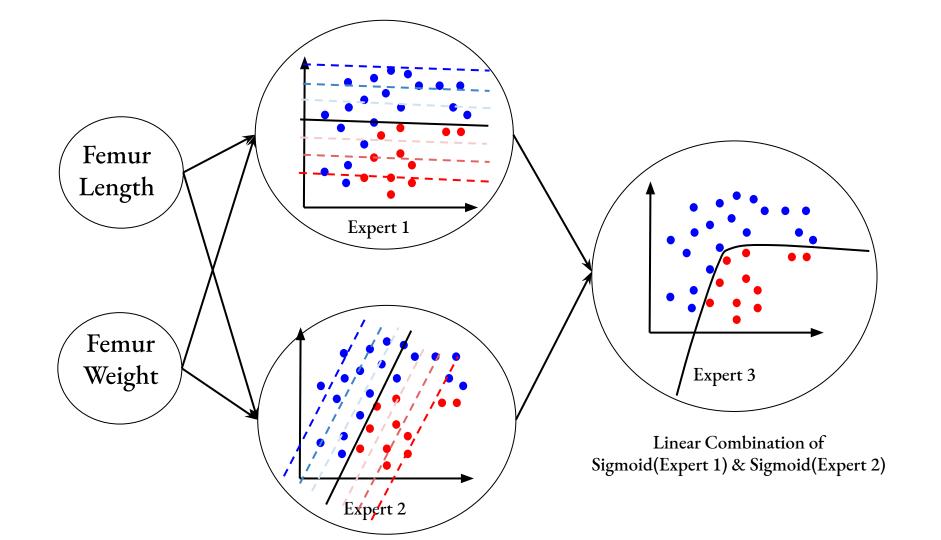


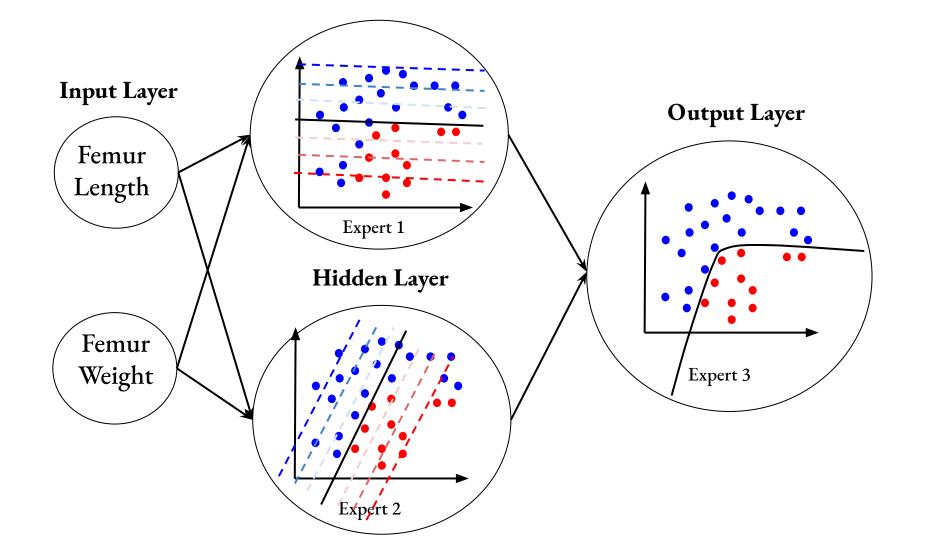
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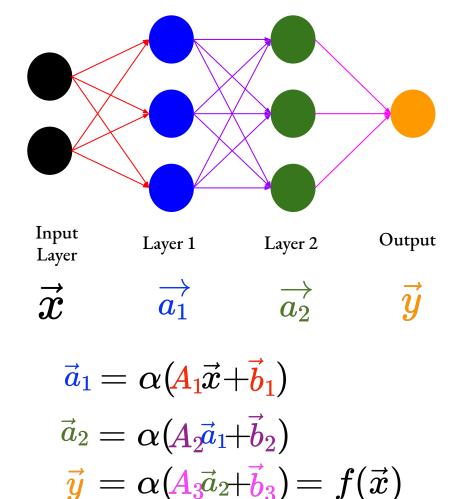
$$lpha[a_1+a_2(length)+a_2(weight)]$$



Sigmoid function





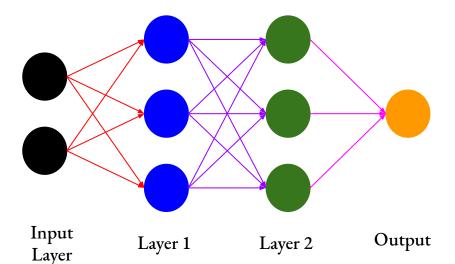


In general:

- Linear function followed by nonlinearity
- Can approximate *any* function this way
- 'Deeper' = more layers

Why?

- Linear = simple, fast
- Optimization



$$\vec{x}$$
 \bar{a}

$$\overrightarrow{a_1}$$

$$\overrightarrow{a_2}$$

$$\overrightarrow{a_2}$$

$$ar{y}$$

Linear Algebra

$$+ \overrightarrow{b}_1)$$

$$egin{align} ec{a}_1 &= lpha ig(\!\! A_1 \! ec{x} \! + \! ec{b}_1 ig) \ ec{a}_2 &= lpha ig(\!\! A_2 \! ec{a}_1 \! + \! ec{b}_2 ig)
onumber \end{aligned}$$

$$ec{oldsymbol{y}}=lpha(\!A_3\!ec{a}_2\!+\!ec{b}_3\!)\!=f(ec{x})$$

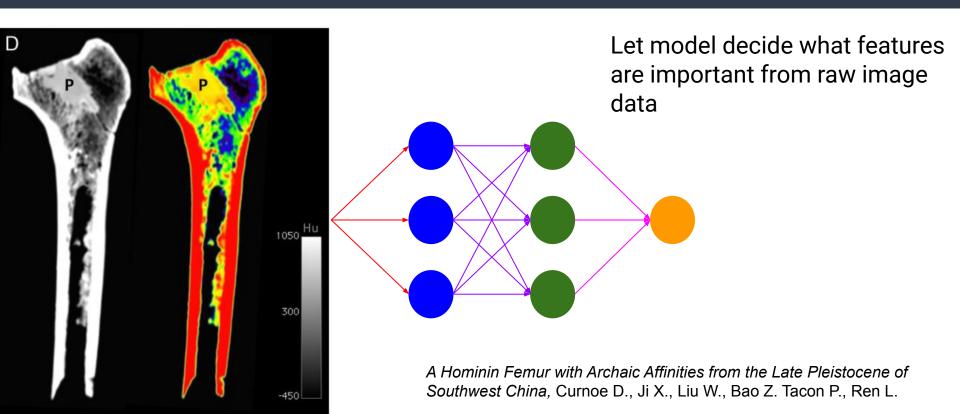
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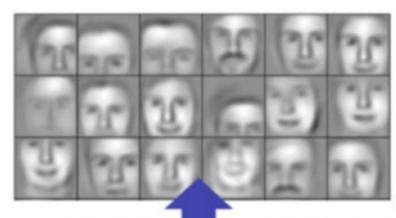
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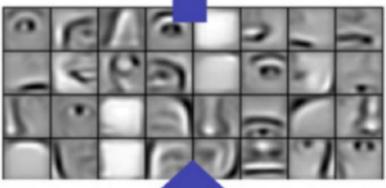
Calculus

- Linear = simple, fast
- **Optimization**





Layer 3



Layer 2



Layer 1 Convolutional Deep Belief Networks for Scalable Unsupervised Laerning of Hierarchical Representations, Lee H., Grosse R., Ranganath R., Ng A.

Why now?

- To build a high-performance Neural Network model you need:
 - Lots of Data
 - Lots of Computing Power
- The sheer amount of data prevents overfitting and helps Neural
 Networks learn complicated patterns in data
- GPUs perform small computations extremely quickly and in parallel, incidentally they are great for training Neural Networks

Benefits of Recent Progress in Deep Learning

- Explosion of performance capability
 - Computer Vision: from fun Snapchat features to self-driving
 - NLP: Conquered simple language tasks, conversation/Q&A is next?
 - Al that can beat humans at games like Go
- Upends many traditional modelling frameworks
- Gives mathematicians/statisticians something to do...

Careful!

- Neural Networks are generally not interpretable.
 - We don't know why/how a particular model comes to its decision
 - Hard to know exactly what it will do in outlier situations
 - Wholly dependent on data it has access to

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- Can be dangerous when used carelessly with real world consequences
 - Can reinforce systemic biases (yes algorithms can learn to be racist)
 - Liability issues for Al

Careful! - Examples

- Man is to Computer Programmer as Woman is to Homemaker? Debiasing
 Word Embeddings by Bolukbasi, Chang, Zou, Saligrama, Kalai
- Examining the Modelling Framework of Crime Hotpot Models in Predictive Policing by Goodson, Hoyer-Leitzel
- Face Recognition: Too Bias or Not Too Bias? by Robinson, Livitz, Henon, Qin, Fu, Timoner

Summary

- Deep learning works by combining simple computations to create a complicated network with many parameters then using tons of data to train a high-performing model
- These algorithms are becoming increasingly common
- Can be extraordinarily helpful
- Can also be extremely harmful if not used carefully and if developers do not think deeply about their impact
- Still largely a "black box" method with more study desperately needed!