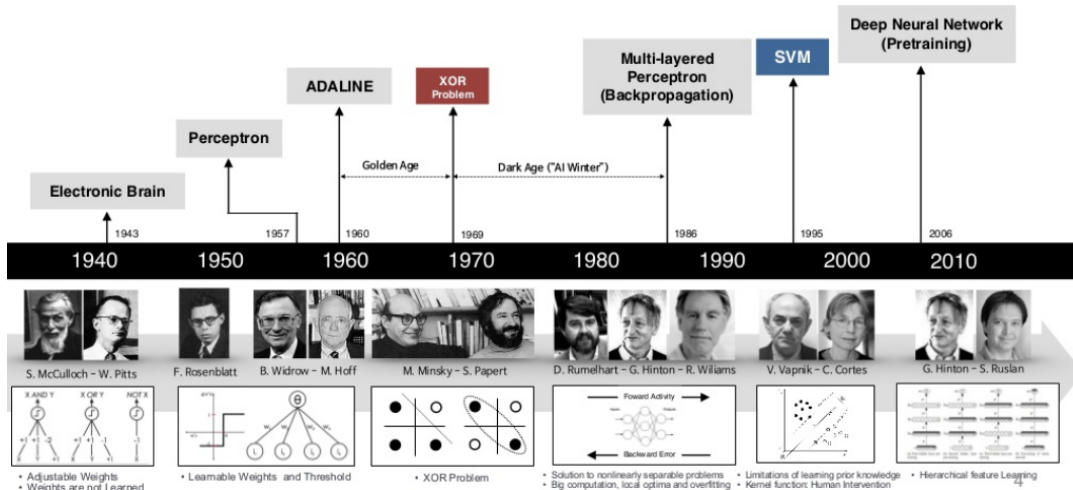


Neural Networks and Deep Learning: Success Deep Learning History

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80's: 1st Convolutional Neural Networks

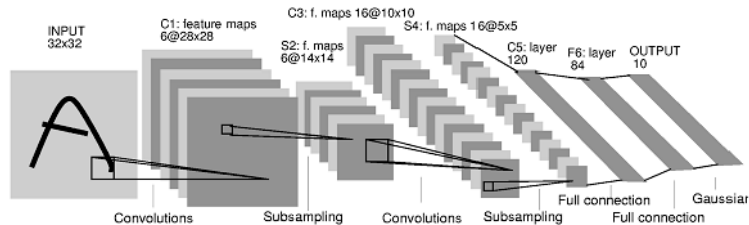


Slide credit:

<https://www.slideshare.net/deview/251-implementing-deep-learning-using-cu-dnn>

Case Study: LeNet 5 Model [LeCun et al., 1989]

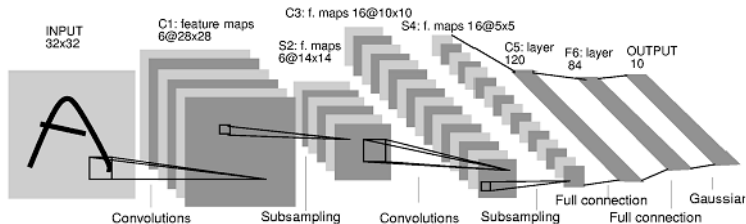
- ▶ **LeNet 5 Model, trained using back-prop**
- ▶ Input: 32x32 pixel image



- ▶ **Macro architecture:** [Convolution/Pooling] blocks C_x + Fully Connected layers F_x
- ▶ C5: convolution layer ~ fully connected

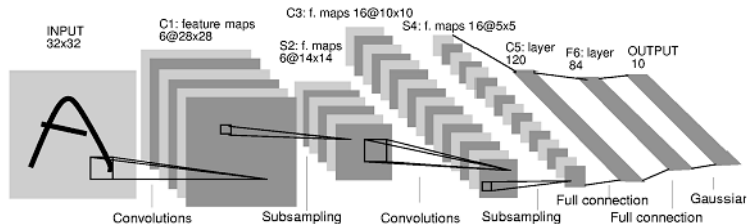
C1 Layer

- ▶ **Input: 32x32x1 tensor (image)**
- ▶ Convolutional layer with 6 5x5 filters
⇒ **Output: 28x28x6 tensor**
- ▶ # Parameters: 5^2 per filter + bias
⇒ $(5 * 5 + 1) * 6 = 156$
 - ▶ If it was fully connected:
 $(32*32+1)*(28*28)*6$ parameters $5 \sim 10^6$!



S2 Layer

- ▶ Subsampling layer = pooling layer

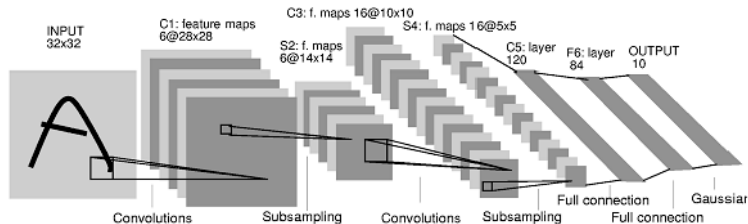


- ▶ **Input: 28x28x6 tensor**
- ▶ Pooling parameters: area (2x2 in C1), stride 2
 - ▶ $p(x) = w \cdot \sum_{i=1}^4 x_i + b \Rightarrow 2 \text{ parameters per channel}$
- ▶ **Output: 14x14x6 tensor**
- ▶ Total # Parameters: $2 * 6 = 12$

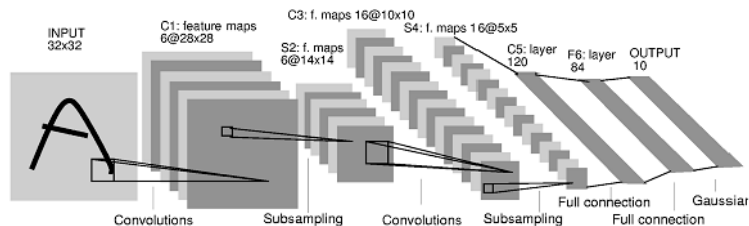
C3 Layer: Convolutional

- ▶ **Input: 14x14x6 tensor**
- ▶ **C3 Convolutional, 16 filters of spatial size 5x5**
 - ▶ **Sparse depth connection to a subset of S2 maps**
⇒ 0-5 connected to 3, 6-14 to 4, 15 connected to 6
 - ▶ **Sparse connection goal?**
- ▶ **Output: 10x10x16 tensor**
- ▶ **# Parameters?**

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	X				X	X	X			X	X	X	X		X	X
1	X	X				X	X	X			X	X	X	X		X
2	X	X	X				X	X	X			X		X	X	X
3		X	X	X			X	X	X	X			X		X	X
4			X	X	X			X	X	X	X		X	X		X
5				X	X	X			X	X	X	X		X	X	X

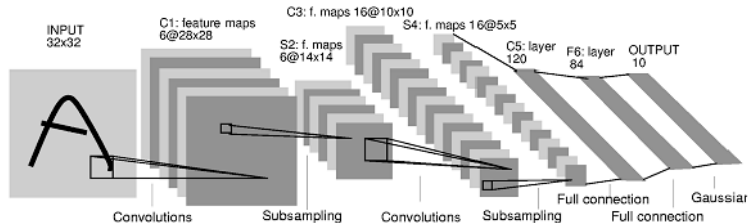


S4 Layer



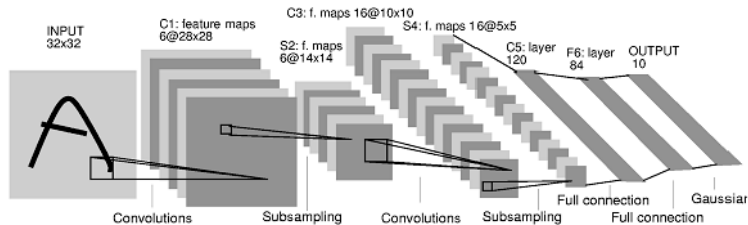
- ▶ Subsampling layer = pooling ~ S2
- ▶ **Input: 10x10x16 tensor**
- ▶ **Output: 5x5x16 tensor**
- ▶ Total # Parameters: $2 * 6 = 12$

C5 Layer: Convolutional



- ▶ **Input: $5 \times 5 \times 16$ tensor**
- ▶ C5 Convolutional, 120 $5 \times 5 \times 16$ filters
⇒ whole depth of S4 (\neq C3)
- ▶ **Output: $1 \times 1 \times 120$ tensor = vector!**
⇒ C5 ~ FC layer on flattened tensor!
- ▶ Total # Parameters?

F6-F7 Layers: Fully Connected

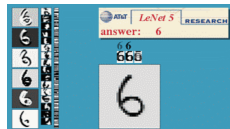
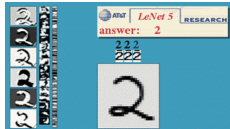
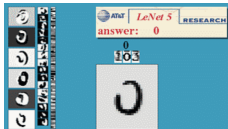


- ▶ F6 Layer
 - ▶ 84 fully connected units.
 - ▶ # Parameters: $84 \times (120 + 1) = 10164$
- ▶ F7 Layer (output): Fully Connected layer
 - ▶ 10 (# classes) fully connected units.
 - ▶ # Parameters: $10 \times (84 + 1) = 850$

LeNet 5 Model: Performances on MNIST

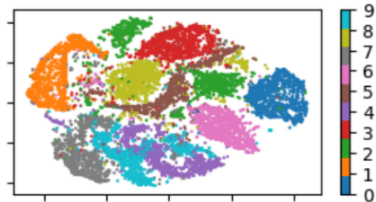
- ▶ Evaluation on MNIST
- ▶ Total # parameters ~ 60000
 - ▶ 60,000 original datasets: test error: 0.95%
 - ▶ + 540,000 artificial distortions \Rightarrow test error: 0.8%
- ▶ **Successful deployment for postal code reading in the US**

3 6 8 1 7 9 6 6 9 1
6 7 5 7 8 6 3 4 8 5
2 1 7 9 7 1 2 8 4 5
4 8 1 9 0 1 8 8 9 4
7 6 1 8 6 4 1 5 6 0
7 5 9 2 6 5 8 1 9 7
2 2 2 2 2 3 4 4 8 0
0 2 3 8 0 7 3 8 5 7
0 1 4 6 4 6 0 2 4 3
7 1 2 8 7 6 9 8 6 1

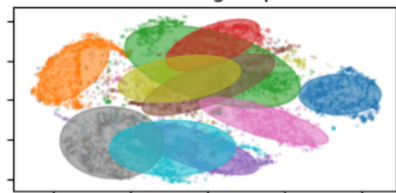


LeNet 5 Model: Manifold Untangling

NH=89.8533333333

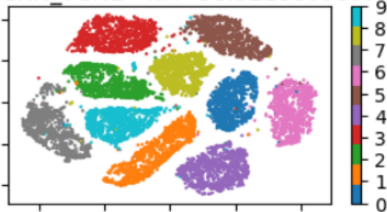


t-SNE fitting ellipses

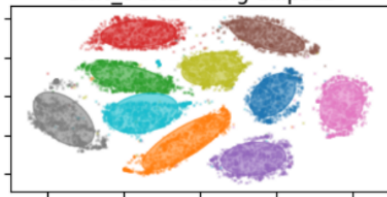


Input space

2D CNN t-SNE - NH=98.5216666667



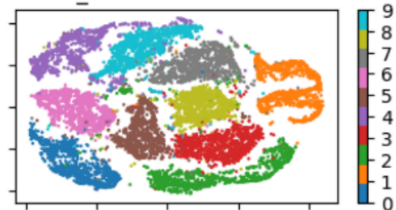
CNN t-SNE fitting ellipses



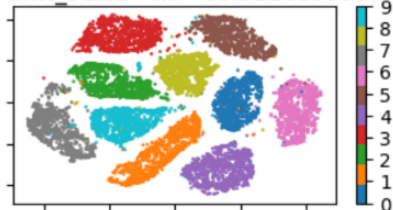
Latent space

LeNet 5 Model: Manifold Untangling

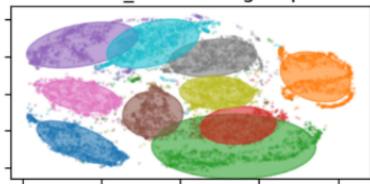
2D MLP100_t-SNE - NH=94.9333333333



2D CNN_t-SNE - NH=98.5216666667



MLP100_t-SNE fitting ellipses



Latent space MLP

CNN_t-SNE fitting ellipses

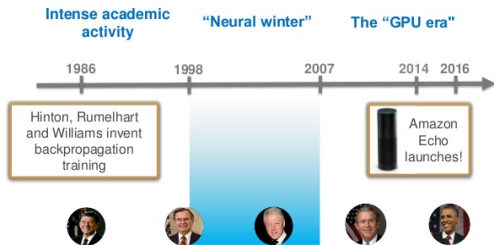


Latent space LeNet

80's: Deep Learning Succeeds

- ▶ ConvNet trainable with backprop, able to untangle class manifold
- ▶ Very good performances for digit classification, industrial transfer
- ▶ **2nd winter of Deep Learning**
⇒ following!

History of Deep Learning



References I



LeCun, Y., Boser, B., Denker, J. S., Henderson, D., Howard, R. E., Hubbard, W., and Jackel, L. D. (1989).
Backpropagation applied to handwritten zip code recognition.
Neural computation, 1(4):541–551.