

# CS540 Introduction to Artificial Intelligence

## Lecture 14

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Based on lecture slides by Jerry Zhu and Yingyu Liang

July 2, 2019





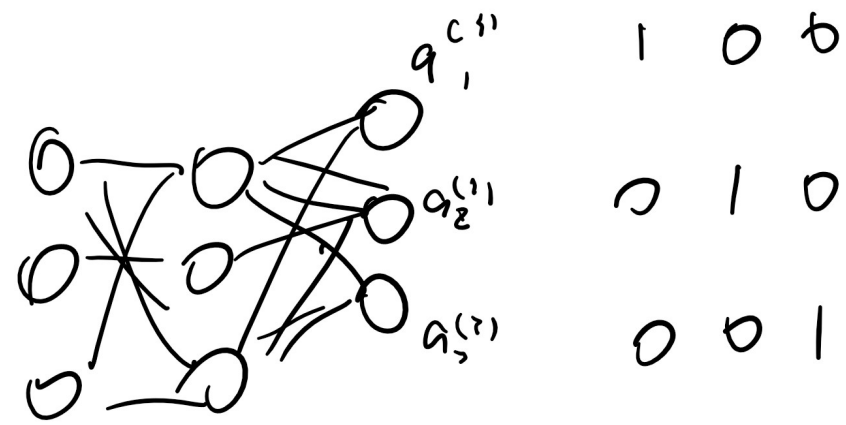
# Neural Network

## Review

- Activation.
- Backpropogation.
- $L_1$  and  $L_2$  regularization.
- Cross validation.
- Multi class classification.

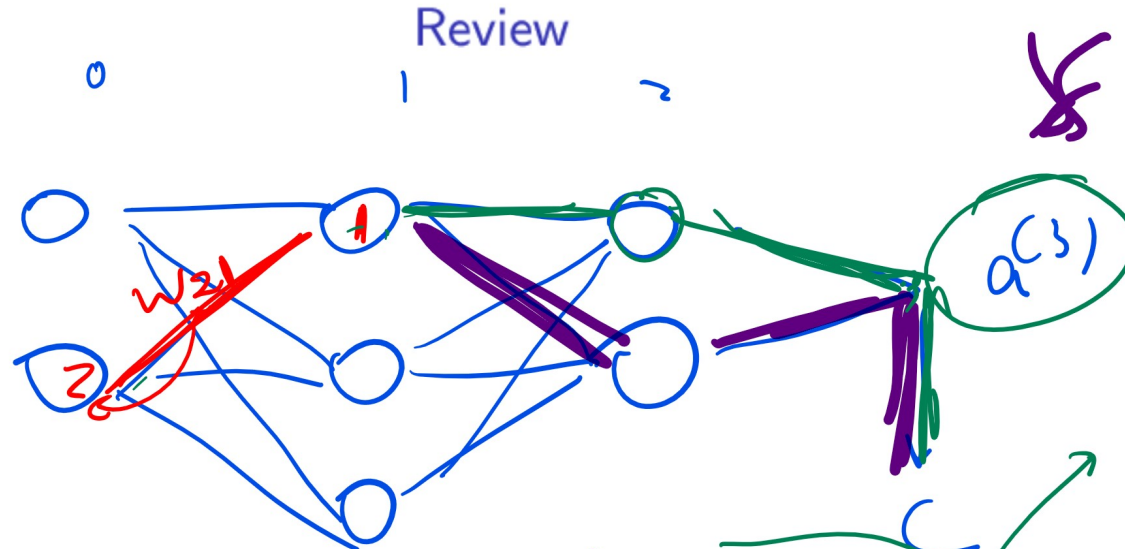
give NN  $\rightarrow$   $\hat{y}$   
 $\frac{\partial C}{\partial w}$

$y = 1, 2, 3$



$$C = \sum_{i=1}^n \left( \frac{1}{2} (y_{i1} - a_{i1}^{(3)})^2 + \frac{1}{2} (y_{i2} - a_{i2}^{(3)})^2 + \frac{1}{2} (y_{i3} - a_{i3}^{(3)})^2 \right)$$

# Multi Layer Neural Network Example

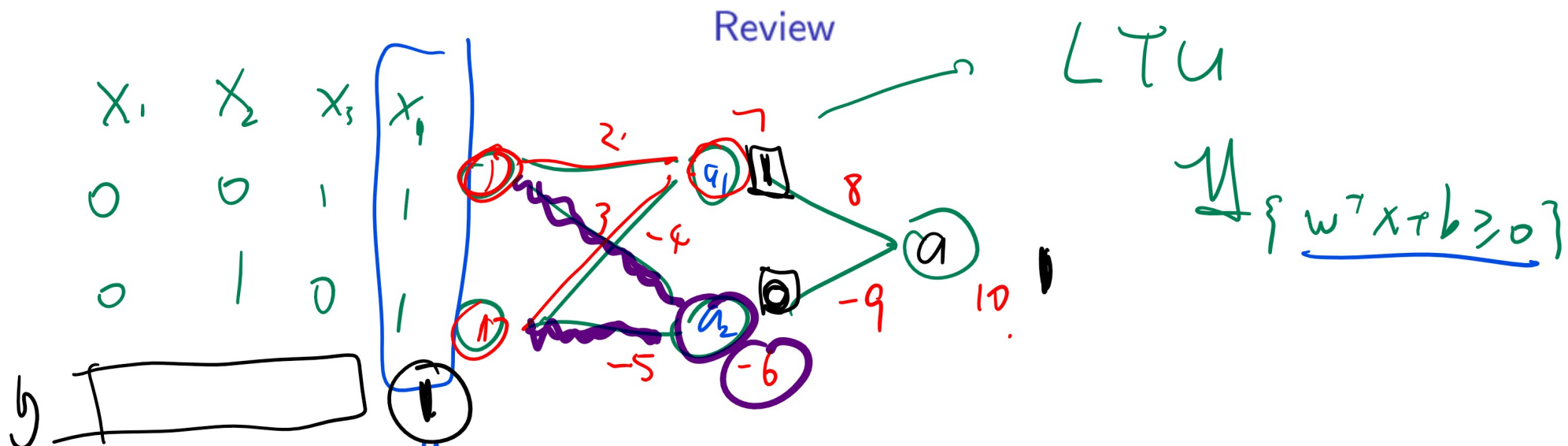


$$\frac{\partial C}{\partial v_{21}} = \frac{\partial C}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial a_2^{(2)}} \cdot \frac{\partial a_2^{(2)}}{\partial a_1^{(1)}} + \frac{\partial C}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial a_1^{(2)}} \cdot \frac{\partial a_1^{(2)}}{\partial a_1^{(1)}} \cdot \frac{\partial a_1^{(1)}}{\partial w_{21}}$$

$$a_2^{(2)} = g(z_2^{(2)})$$

$$z_2^{(2)} = w_{12} a_1^{(1)} + \dots$$

# LTU Activation Example



$$a_1 = \Downarrow \{1 \cdot 2 + 1 \cdot 3 + 7 \geq 0\} = 1$$

$$a_2 = \Downarrow \{1 \cdot -4 + 1 \cdot -5 - 6 \geq 0\} = 0$$

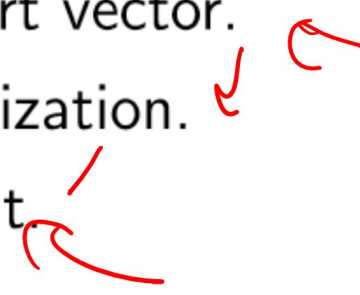
False

$$a = \Downarrow \{1 \cdot 8 + 0 \cdot (-9) + 10 \geq 0\} = 1$$

$\geq 0$

# Support Vector Machine

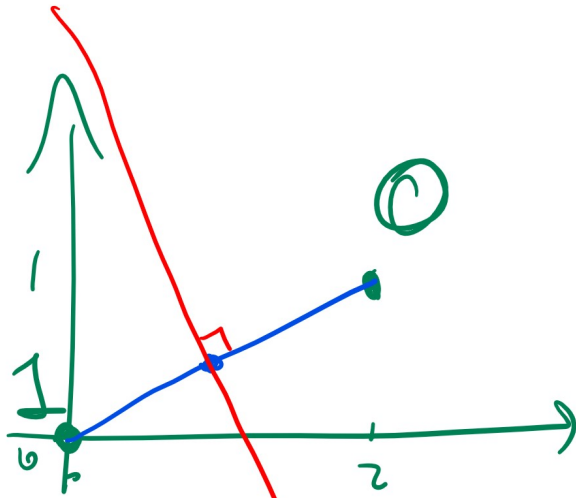
## Review

- Hard margin support vector.
  - Soft margin maximization.
  - Subgradient descent
  - Kernel trick.
- 



## Support Vector Margin Example

Review

slope  $\frac{1}{2}$ SVM slope  $-2$ point  $(1, \frac{1}{2})$ 

$$x_2 = -2x_1 + b$$

$$\frac{1}{2} = -2 \cdot 1 + b \Rightarrow b = \frac{5}{2}$$

SVM



$$\{ 2x_1 + x_2 - \frac{5}{2} \leq 0 \}$$

$$= \{ \frac{4}{5}x_1 + \frac{2}{5}x_2 - 1 \leq 0 \}$$

$$= \{ -\frac{4}{5}x_1 - \frac{2}{5}x_2 + 1 \geq 0 \}$$



# Feature Vector to Kernel Example

Review

~~kernel~~  $\phi(x_1, x_2) = (x_1^2, \sqrt{2}x_1x_2, x_2^2)$

kernel  $\phi^T(\underbrace{x_1, x_2}_{\text{instance 1}}) \phi(\underbrace{x'_1, x'_2}_{\text{instance 2}}) = (x_1^2, \sqrt{2}x_1x_2, x_2^2) \begin{pmatrix} x_1^2 \\ \sqrt{2}x_1x_2 \\ x_2^2 \end{pmatrix}$

instances

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$K = \begin{pmatrix} \boxed{1} & \boxed{0} \\ \boxed{0} & \boxed{1} \end{pmatrix}$$

$$= (x_1x'_1)^2 + 2(x_1x'_1x_2x'_2) + (x_2x'_2)^2$$

$$= (x_1x'_1 + x_2x'_2)^2$$

$$x = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad x' = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$x = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad x' = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

# Decision Tree

## Review

- Entropy.
- Information gain.
- Bagging and boosting.

# Decision Tree Example

NXOR Review

	$x_1$	$x_2$	$y$
①	0	0	1
②	0	1	0
③	1	0	0
④	1	1	1

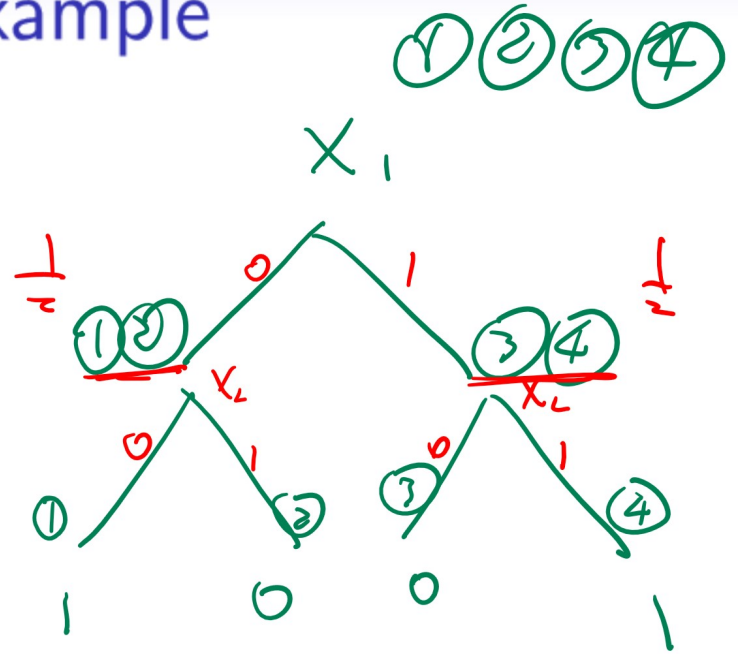
$H(Y | x_1) = 1$

$$\frac{1}{2} H(Y | x_1 = 0) + \frac{1}{2} H(Y | x_1 = 1)$$

entropy of ①②  
 $\frac{1}{2} \quad \frac{1}{2}$

$$-\left( \frac{1}{2} \log_2\left(\frac{1}{2}\right) + \frac{1}{2} \log_2\left(\frac{1}{2}\right) \right)$$

+ 1



$$\frac{1}{2} H(Y | x_1 = 1)$$

entropy of ③④

$$+ 1$$

$H(Y) = 1$

$I_G = 0$

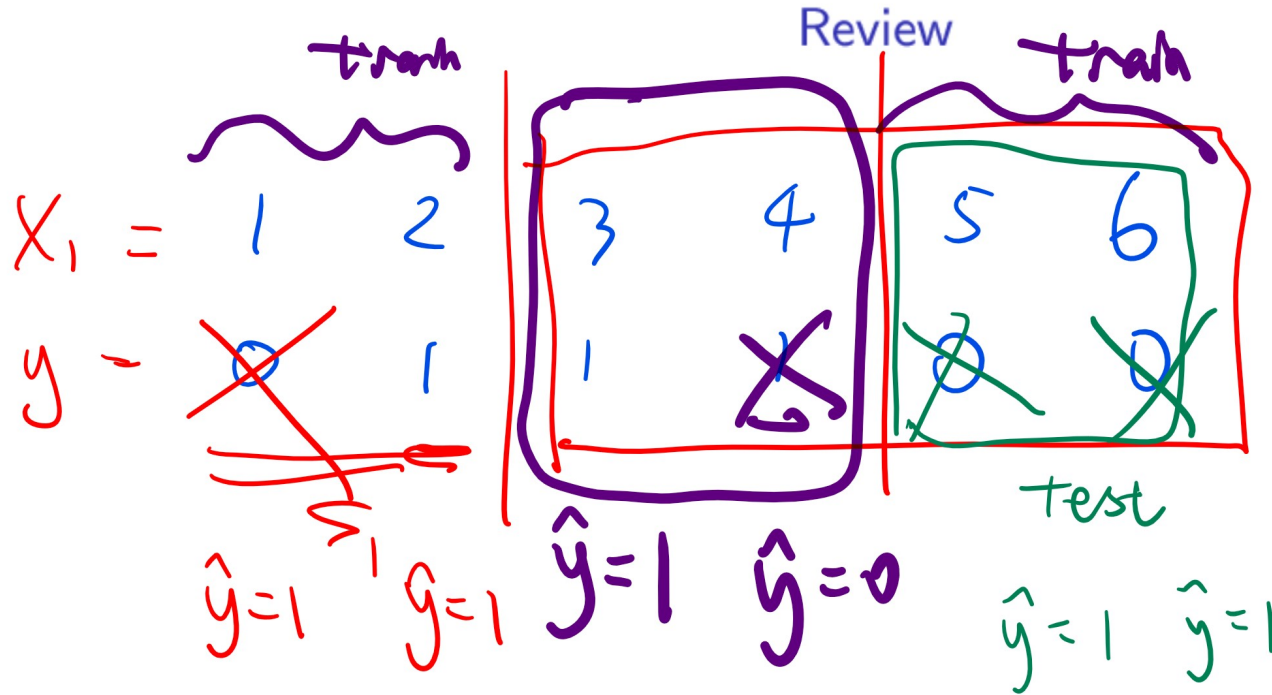
# K Nearest Neighbor

## Review

- Distance functions.

$L_1, L_2, L_\infty$

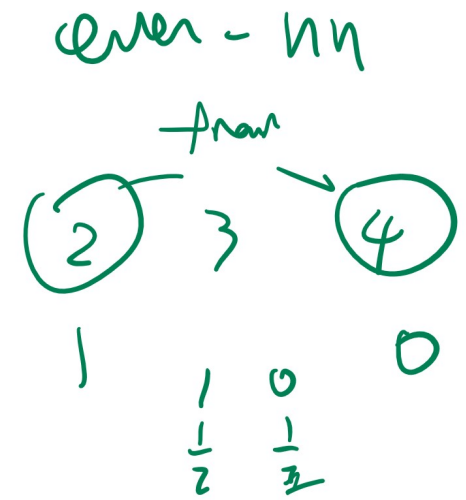
# K Nearest Neighbor Cross Validation Example



1 - nn

3-fold CV.

$$\text{Accuracy} = \frac{2}{6} = \frac{1}{3}$$

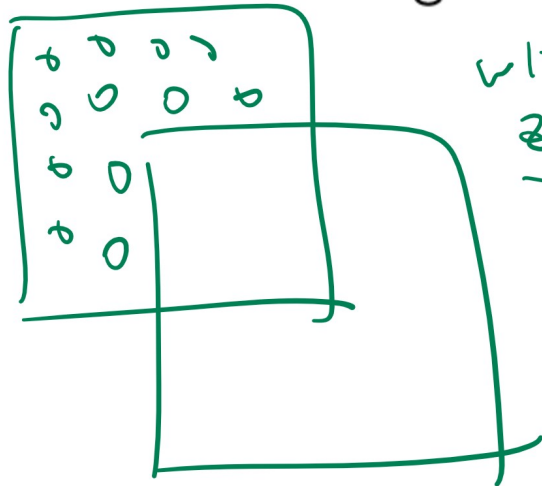


# Convolutional Neural Network

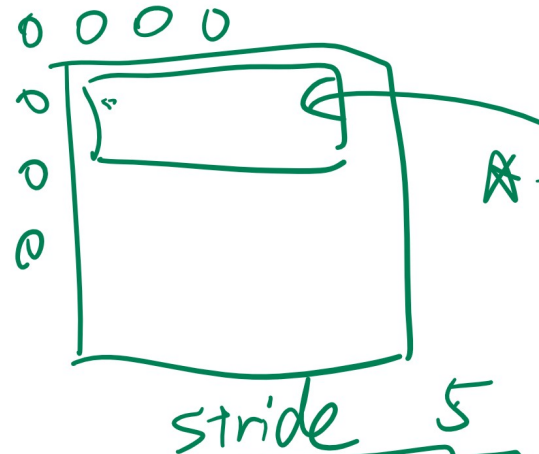
## Review

flip

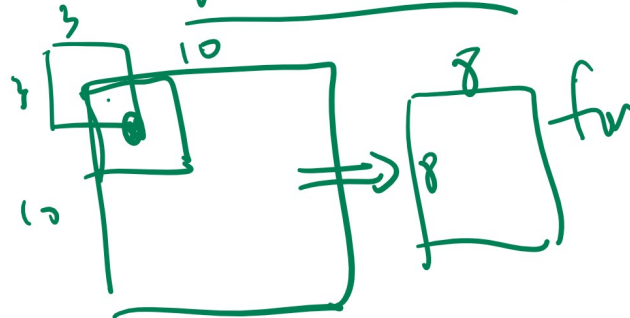
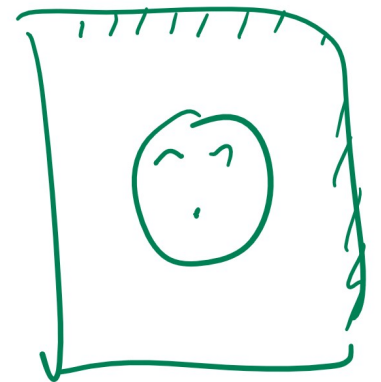
- Convolution.
- Pooling.
- Trained weights.



without zero padding



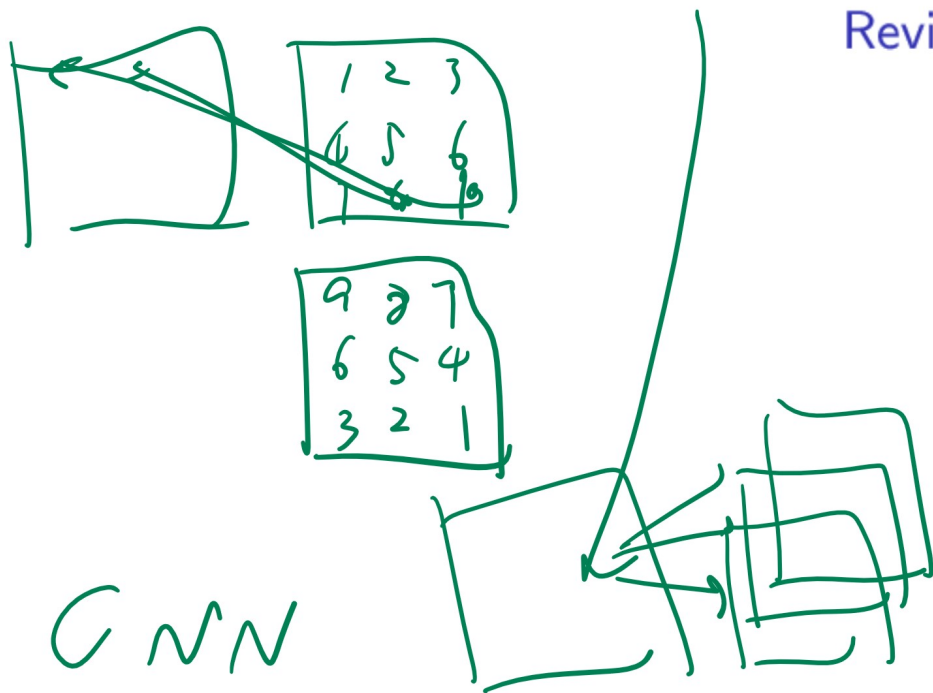
Stride 1



for HW 4  
 C on boundary is 0

# Convolutional Weights Count Example

Review



CNN



filters Conv  
pool layer

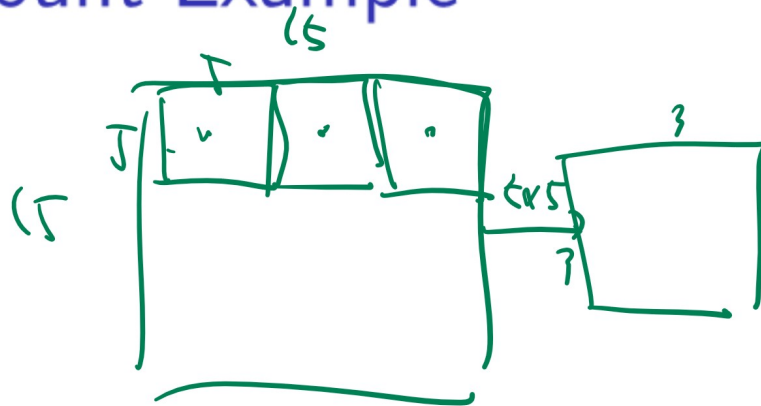
nonoverlapping + stride 5

all need trained  
none

5x5 filter

train 25 weights.

fully connected all need trained



15x15 activation map matrix

5x5 pooling filter

3x3 activation map.

# Computer Vision

## Review

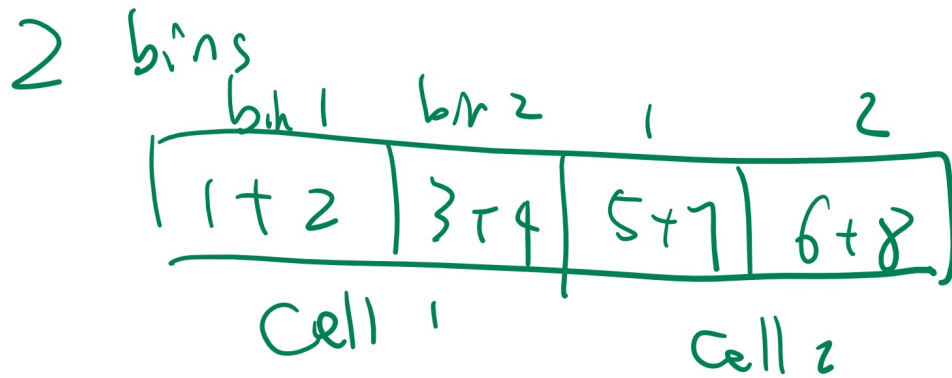
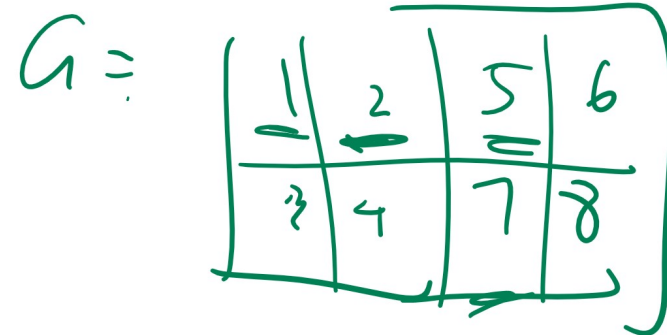
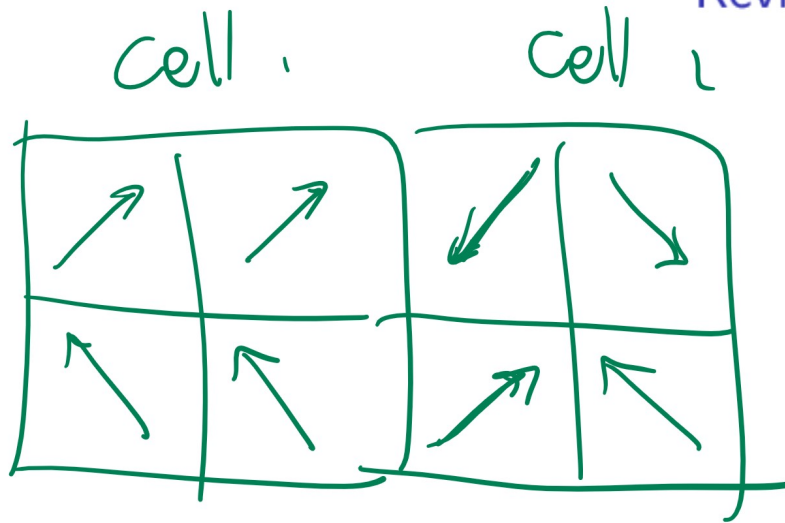
- Histogram of Gradients Features.
- Scale Invariant Feature Transform.
- ~~Block normalization.~~
- Dominant orientation.
- Harr Features.

give his-gram



# Histogram of Gradient Example

Review



# Natural Language Processing

## Review

- Bigram and trigram model.
- Transition matrix.
- Random word generation.
- Bayes rule.

$$Pr \{ \underline{w} \mid \underline{a} \} = \frac{C_{aw}}{C_a}$$

	a	b	c
a	0.8	0.1	0.1
b			
c			

Laplace  
smooth

a b

$$\frac{C_{aw} + 1}{C_a + |\text{vocabulary}|}$$

## Document Bayes Rule Example

Review

$$\frac{1}{3} \begin{bmatrix} A \rightarrow 0.7 & H \\ B \rightarrow 0.5 & H \\ C \rightarrow 0.2 & H \end{bmatrix}$$

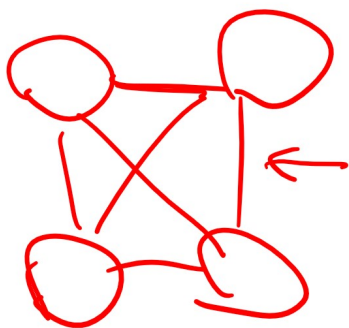
$$P_r(B | H) = \frac{P_r(B|H)}{P_r(H)} \Rightarrow \frac{P_r(H|B) P_r(B)}{\frac{1}{2} \cdot \frac{1}{3}}$$

$$\Rightarrow \frac{1}{3} \cdot 0.7 + \frac{1}{3} \cdot 0.5 + \frac{1}{3} \cdot 0.2$$

# Bayesian Network

## Review

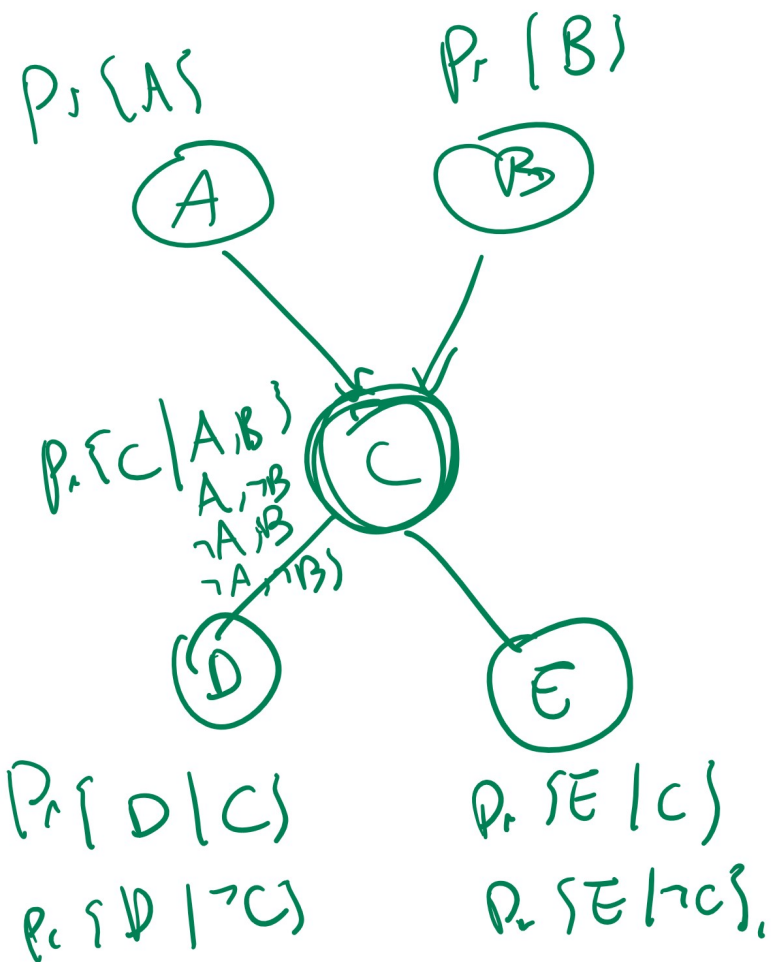
- Conditional probability table.
- Maximum likelihood estimation.
- Training vs inference.
- Chow Liu algorithm.



max  
Sparsity  
free

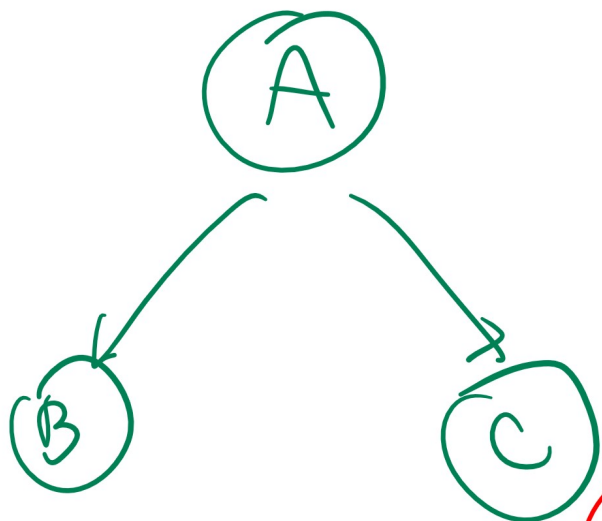
$$\frac{C_{C=1, D=1} + 1}{C_{C=1} + 2} \approx$$

$$\frac{C_{C=0, D=1} + 1}{C_{C=0} + 2} \approx$$



# Common Cause Network Example

Review



$$P_r(A) = 0.7$$

$$P_r(B|A) = 0.3$$

$$P_r(B|\neg A) = 0.2$$

$$P_r(C|A) = 0.4$$

$$P_r(C|\neg A) = 0.6$$

$$P_r(B|\neg C) = \frac{P_r(B, \neg C, A=0)}{P_r(\neg C)}$$

$$P_r(B, \neg C, A) + P_r(B, \neg C, \neg A)$$

$$P_r(B|A) \cdot P_r(\neg C|A) \cdot P_r(A) + P_r(B|\neg A) \cdot P_r(\neg C|\neg A) \cdot P_r(\neg A)$$

$$P_r(B, \neg C) = 0.3 \cdot 0.6 \cdot 0.7 + 0.2 \cdot 0.4 \cdot 0.3$$

$$\begin{aligned} P_r(\neg C) &= P_r(\neg C, A) + P_r(\neg C, \neg A) \\ &= P_r(\neg C|A) \cdot P_r(A) + P_r(\neg C|\neg A) \cdot P_r(\neg A) \end{aligned}$$

$$P_r(\neg C) = 0.6 \cdot 0.7 + 0.4 \cdot 0.3$$