

CS540 Introduction to Artificial Intelligence

Lecture 14

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Based on lecture slides by Jerry Zhu and Yingyu Liang

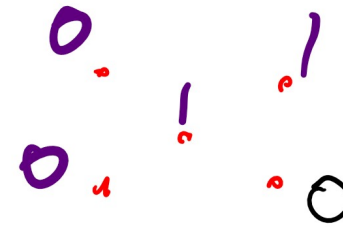
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Perceptron

Review

- Perceptron update rule.
- Perceptron termination condition.

 iff data points linearly separable

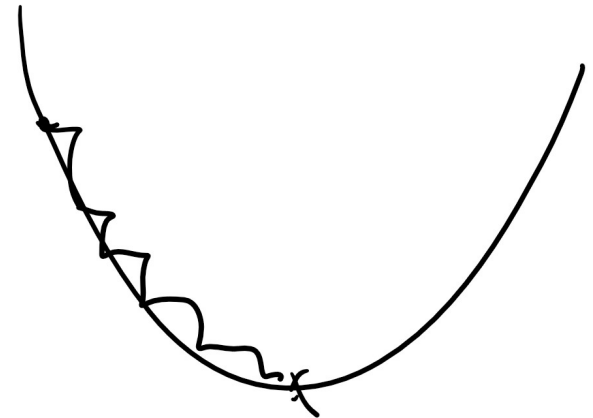


Logistic Regression

Review

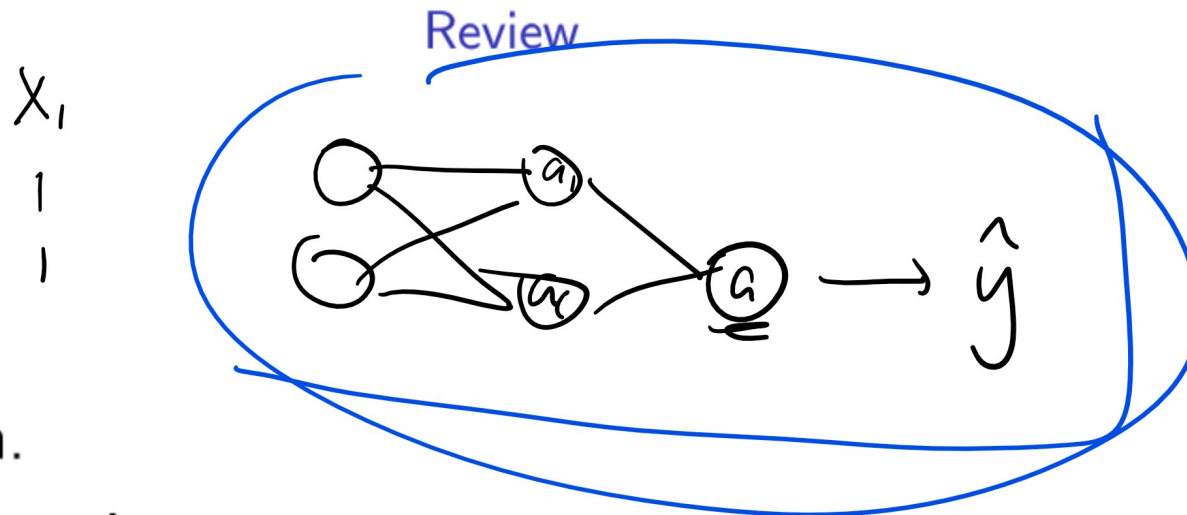
- Logistic update rule. ↙
- Logistic cost function. → log cost.
- Convexity. ←
- Hessian, Laplacian, eigenvalue. ↘

↙ eigenvalue ≥ 0



$$\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

Neural Network



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

One vs one

one vs all

softmax

$y = \frac{1}{z} \Rightarrow$

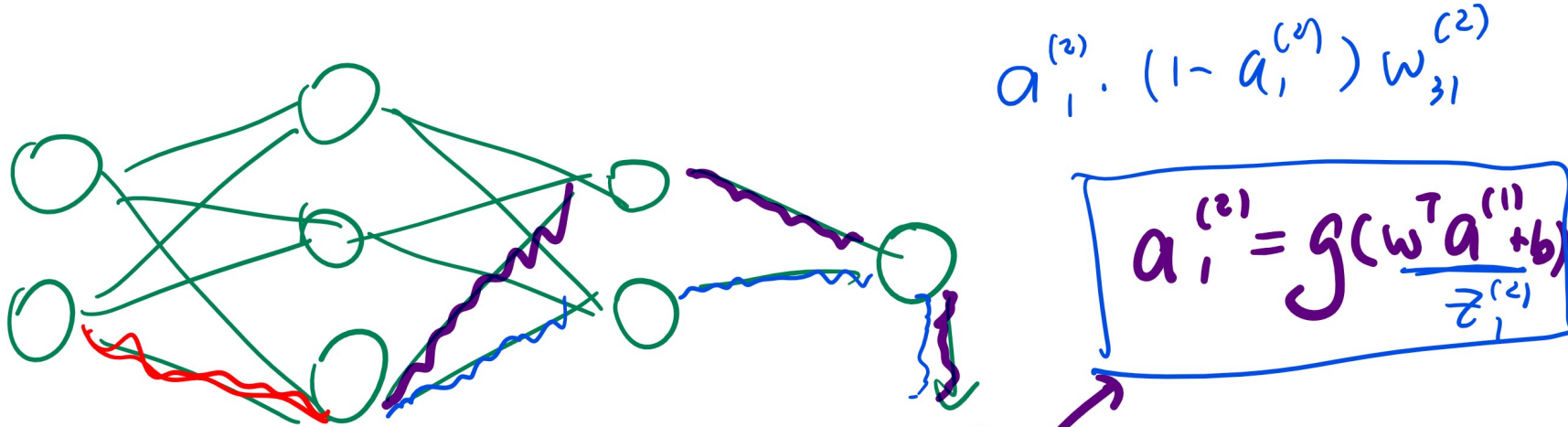
3 activation in last layer

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

$$C = \sum_{i=1}^n (y_i - a_i^{(s)})^2$$

Multi Layer Neural Network Example

Review



$$\frac{\partial C}{\partial w_{23}} = \frac{\partial C}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial a^{(2)}} \cdot \frac{\partial a^{(2)}}{\partial a^{(1)}} \cdot \frac{\partial a^{(1)}}{\partial w_{23}}$$

$$\rightarrow \frac{\partial C}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial a_2^{(2)}} \cdot \frac{\partial a^{(2)}}{\partial a_3^{(1)}} \cdot \frac{\partial a^{(1)}}{\partial w_{23}}$$

LTU Activation Example

Review

Hw 2

Support Vector Machine

Review

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

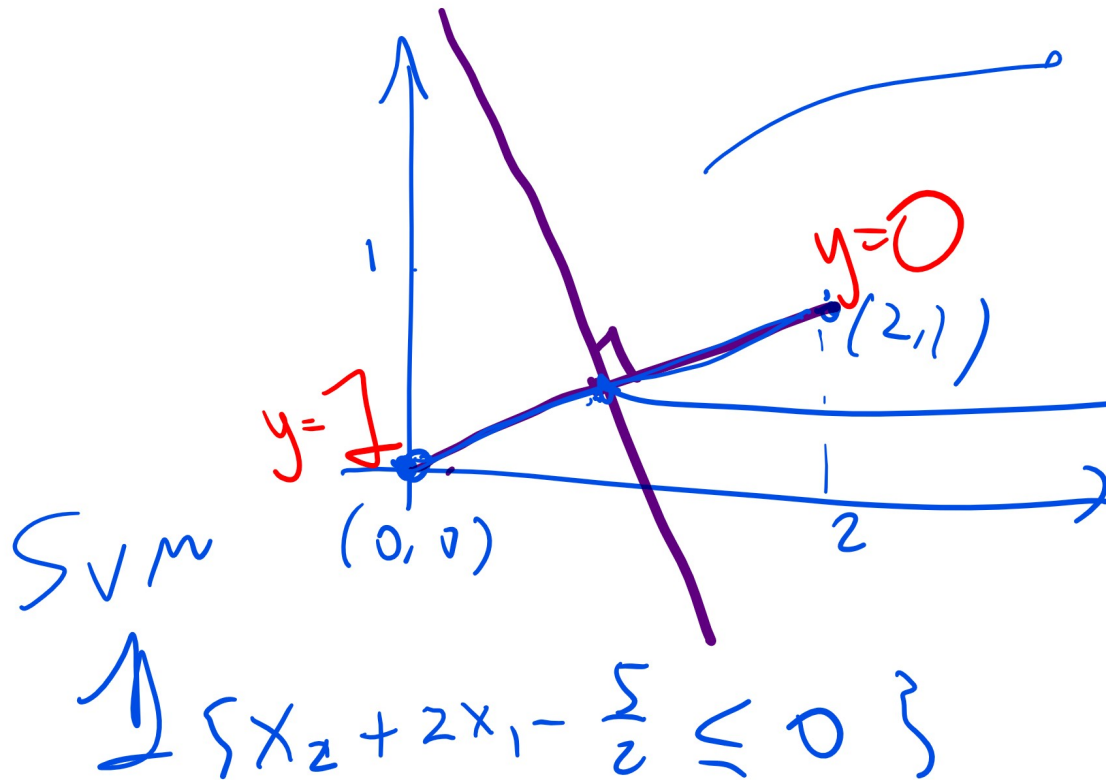
Support Vector Margin Example

Review

$$\text{slope} = -\frac{1}{2}$$

$$= -2$$

$$\text{point} = \left(1, \frac{1}{2}\right)$$



$$x_2 = -2x_1 + b$$

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

$$b = \frac{5}{2}$$

Feature Vector to Kernel Example

Review

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

$$K = \phi^T(\overset{x}{x_1, x_2}) \phi(\overset{x'}{x'_1, x'_2}) = (x_1^2, \sqrt{2} x_1 x_2, x_2^2) \begin{pmatrix} x_1'^2 \\ \sqrt{2} x_1' x_2' \\ x_2'^2 \end{pmatrix}$$

$$= (x_1 x_1')^2 + 2(x_1 x_1')(x_2 x_2') + (x_2 x_2')^2$$

$$= (x_1 x_1' + x_2 x_2')^2$$

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

$$K(x, x') = 0$$

Decision Tree

Review

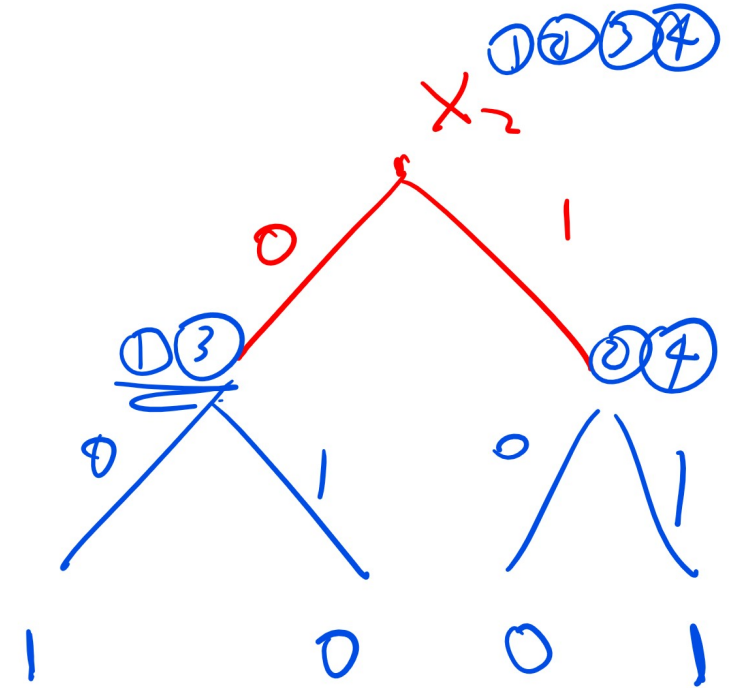
- Entropy. ↩
- Information gain.
- Bagging and boosting. ↩

Decision Tree Example

Review
N XOR

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

$\frac{1}{2}$, $\frac{1}{2}$

 $H(Y)$

$$\begin{aligned}
 H(Y | X_2) &= P_{\{X_2=0\}} \cdot H(Y | X_2=0) + P_{\{X_2=1\}} \cdot H(Y | X_2=1) \\
 &= \frac{1}{2} \left(\underbrace{-\frac{1}{2} \log \frac{1}{2}}_{Y=0} - \underbrace{\frac{1}{2} \log \frac{1}{2}}_{Y=1} \right) + \frac{1}{2} (1) \\
 &= \left(\frac{1}{2} + \frac{1}{2} \right) = 1 = 1
 \end{aligned}$$

K Nearest Neighbor

Review

- Distance functions.

L_1, L_2, L_∞

K Nearest Neighbor Cross Validation Example

Review

	S_1		S_2		S_3	
$X_i =$	1	2	3	4	5	6
$y_i =$	0	0	1	1	1	0
			3.5	1/2	1	

train on S_2, S_3 , test on $S_1 \Rightarrow$

S_2
 S_3

}

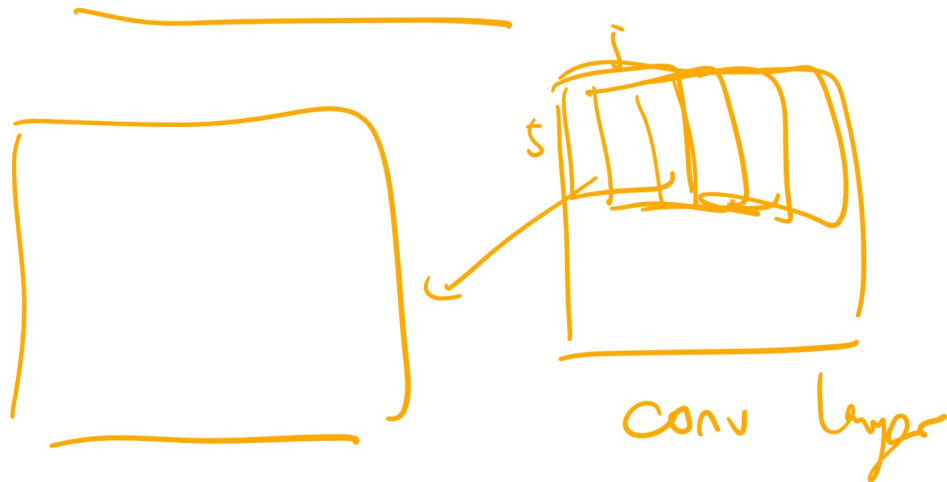
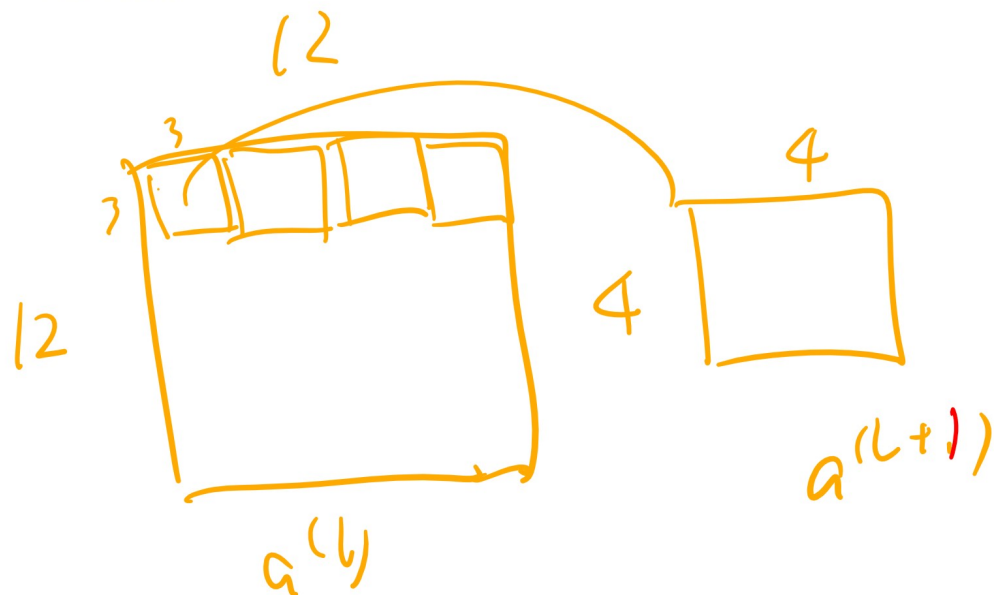
accuracy = $\frac{2}{6} = \frac{2}{3}$

1- nn.
3-fold CV

Convolutional Neural Network

Review

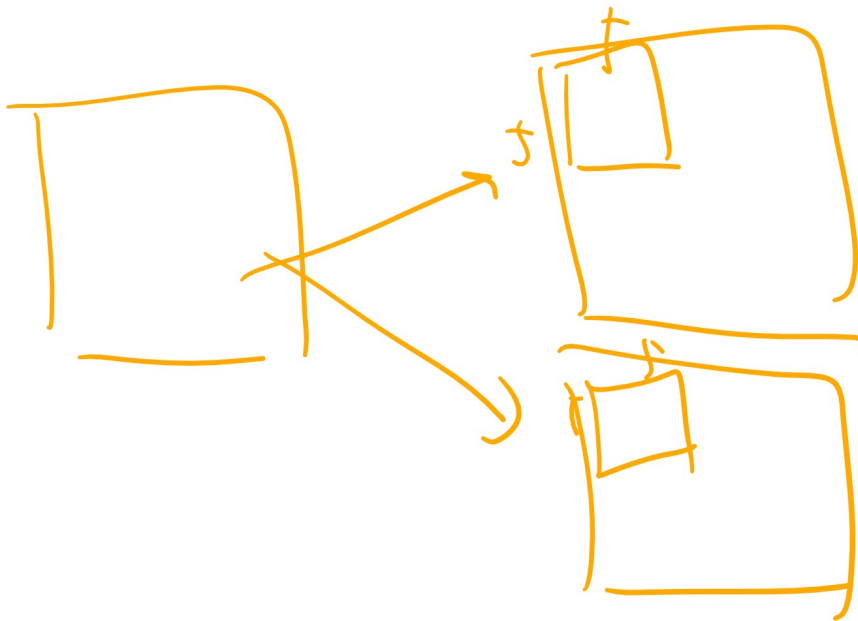
- Convolution.
- Pooling.
- Trained weights.



trained weights in conv layer
 = # elements of filters
 = 25
 # weights in pooling = 1

Convolutional Weights Count Example

Review



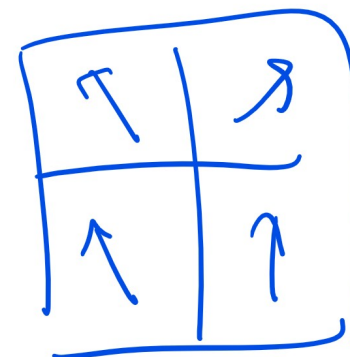
$$5 \times 5 \times 2 = 50$$

Computer Vision

Review

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

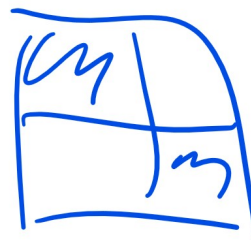
← HWG



add up G

not counting

←



Histogram of Gradient Example

Review

Natural Language Processing

Review

HW5

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

$P_s(a|b)$

$$\frac{C_{ba} + 1}{C_b + |\text{vocabulary}|}$$

$a b b b c \rightarrow ?$

Document Bayes Rule Example

Review

A	70%	"the"	}	$\frac{1}{3}$
B	20%	"the"		$\frac{1}{3}$
C	10%	"the"		$\frac{1}{3}$

$$Pr(B | \text{"the"}) =$$

$$\frac{Pr(B, \text{"the"})}{Pr(\text{"the"})} \rightarrow \frac{Pr(B) \cdot Pr(\text{"the"}|B)}{Pr(\text{"the"})}$$

$$= \frac{\frac{1}{3} \cdot 0.2}{\frac{1}{3} \cdot 0.7 + \frac{1}{3} \cdot 0.2 + \frac{1}{3} \cdot 0.1}$$

$$= \frac{Pr(\text{"the"}|A) \cdot Pr(A)}{Pr(\text{"the"}|A) \cdot Pr(A) + Pr(\text{"the"}|B) \cdot Pr(B) + Pr(\text{"the"}|C) \cdot Pr(C)}$$

Bayesian Network

Review

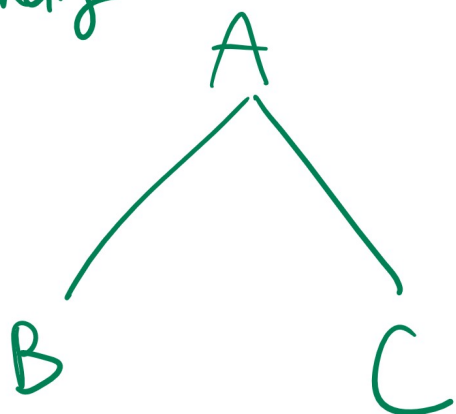
- Conditional probability table.
- Maximum likelihood estimation.
- Training vs inference.
- Chow Liu algorithm.

→ max spanning tree.

$$\frac{C_{AB}}{C_B}$$

Common Cause Network Example

A, B, C binary
0.1



Review

CPT

$$P_r \{A\} = \underline{0.1}$$

$$P_r \{B|A\} = \underline{0.2}, \quad P_r \{\neg B|A\} = 0.8$$

$$P_r \{B|\neg A\} = \underline{0.3}$$

$$P_r \{C|A\} = 0.4$$

$$P_r \{C|\neg A\} = \underline{0.5}$$

$$P_r \{B|\neg C\}$$

=

$$P_r \{B, \neg C, A=0\}$$

$$P_r \{\neg C, A=0\}$$

$$P_r \{B, \neg C, A\} + P_r \{B, \neg C, \neg A\}$$

$$= P_r \{B|A\} \cdot P_r \{\neg C|A\} \cdot P_r \{A\}$$

$$+ P_r \{B|\neg A\} \cdot P_r \{\neg C|\neg A\} \cdot P_r \{\neg A\}$$

$$0.2 \cdot (1 - 0.4) \cdot 0.1$$

$$+ 0.3 \cdot (1 - 0.5) \cdot (1 - 0.1)$$

$$P_1(\neg C, A) + P_2(\neg C, \neg A)$$

$$= (1 - 0.4) \cdot 0.1 + (1 - 0.5) \cdot (1 - 0.1)$$