

UNIVERSITY OF WISCONSIN – MADISON

CS540 – 1: Homework 1

Questions 1 through 3

Matthew Klebenow

9/26/2012

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Question 1: Warm Up Math Questions

Part A:

There are 8 tokens in a bag. 5 are red, 2 are blue, and 1 is green. Without watching, you randomly grab 3 tokens from the bag. What is the chance that you end up with at least 1 red token and at least 1 blue token?

Essentially we have three attempts to satisfy two conditions. As such, the "negative" attempt will not satisfy any remaining criteria for a successful grab. The "negative" can occur as either the first, second, or third grab, and creates six possible scenarios for success. Scenarios 1 & 2 will depict the negative grab third, 3 & 4 will depict the negative grab first, and 5 & 6 will depict the negative grab as second.

Scenario	First Grab	Second Grab	Third Grab	Total Probability	NEGATIVE here is:
RED/BLUE/NEGATIVE	$\frac{5}{8}$	$\frac{2}{7}$	$\frac{6}{6}$	$\frac{5}{8} * \frac{2}{7} * \frac{6}{6} = \frac{5}{28}$	Any color
BLUE/RED/NEGATIVE	$\frac{2}{8}$	$\frac{5}{7}$	$\frac{6}{6}$	$\frac{2}{8} * \frac{5}{7} * \frac{6}{6} = \frac{5}{28}$	Any color
NEGATIVE/RED/BLUE	$\frac{1}{8}$	$\frac{5}{7}$	$\frac{2}{6}$	$\frac{1}{8} * \frac{5}{7} * \frac{2}{6} = \frac{5}{168}$	Neither red nor blue
NEGATIVE/BLUE/RED	$\frac{1}{8}$	$\frac{2}{7}$	$\frac{5}{6}$	$\frac{1}{8} * \frac{2}{7} * \frac{5}{6} = \frac{5}{168}$	Neither red nor blue
RED/NEGATIVE/BLUE	$\frac{5}{8}$	$\frac{5}{7}$	$\frac{2}{6}$	$\frac{5}{8} * \frac{5}{7} * \frac{2}{6} = \frac{25}{168}$	Not blue
BLUE/NEGATIVE/RED	$\frac{2}{8}$	$\frac{2}{7}$	$\frac{5}{6}$	$\frac{2}{8} * \frac{2}{7} * \frac{5}{6} = \frac{5}{84}$	Not red

The total possibility of success is the intersection of all six scenarios. Thus, we add their individual respective probabilities.

$$\sum_{i=1}^6 \text{Scenario}_i = \frac{5}{28} + \frac{5}{28} + \frac{5}{168} + \frac{5}{168} + \frac{25}{168} + \frac{5}{84} = \frac{5}{8}$$

ANSWER:

The probability is $\frac{5}{8}$

Part B:

Take the derivative with respect to x: $(e^x + \sin(x))^3 + x \ln(x)$

Take the derivative with respect to x:

$$\frac{d}{dx}((e^x + \sin(x))^3 + x * \ln(x))$$

We will take the derivative of each term individually.

First term:

$$\begin{aligned} & \frac{d}{dx}(e^x + \sin(x))^3 \\ & \text{let } u = e^x + \sin(x) \\ & \frac{d}{du}u^3 * \frac{du}{dx} \\ & 3u^2 * \frac{d}{dx}(e^x + \sin(x)) \\ & 3(e^x + \sin(x))^2 * \left(\frac{d}{dx}(e^x) + \frac{d}{dx}(\sin(x)) \right) \\ & 3(e^x + \sin(x))^2 * (e^x + \cos(x)) \end{aligned}$$

Second Term:

$$\begin{aligned} & \frac{d}{dx}(x * \ln(x)) \\ & x * \frac{d}{dx}(\ln(x)) + \ln(x) * \frac{d}{dx}(x) \\ & x * \frac{1}{x} + \ln(x) * 1 \\ & 1 + \ln(x) \end{aligned}$$

Combine Terms:

ANSWER:

$$3(e^x + \sin(x))^2 * (e^x + \cos(x)) + \ln(x) + 1$$

Part C:

$$\text{Compute } \lim_{x \rightarrow 1} \frac{\ln(x)}{x - 1}$$

A direct approach of substituting $x = 1$ into the equation yields an indeterminate solution $\left(\frac{0}{0}\right)$. Thus, we must apply L'Hospital's rule.

$$\begin{aligned} & \lim_{x \rightarrow 1} \frac{\frac{d}{dx} \ln(x)}{\frac{d}{dx} (x - 1)} \\ & \lim_{x \rightarrow 1} \frac{1}{1} \end{aligned}$$

ANSWER:

$$\lim_{x \rightarrow 1} \frac{\ln(x)}{x - 1} = 1$$

Question 2: Hierarchical Clustering

Part A:

Given the dataset with five points (0,0),(2,0),(5,0),(0,4),(4,4), run complete-linkage hierarchical clustering by hand. Use L1 distance. For each iteration, show the cluster membership of each point.

Clusters will be represented by placing points within curly braces { }.

Initially, all points are in their own separate clusters.

Depiction of initial state:

D				E	
A		B			C

First iteration:

Cluster	{(0,0)}	{(2,0)}	{(5,0)}	{(0,4)}	{(4,4)}
Min. Distance	2 to {(2,0)}	2, no update	3, no update	4, no update	4, no update

Update Clusters

Cluster	{(0,0), (2,0)}	{(5,0)}	{(0,4)}	{(4,4)}
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Depiction

D				E	
A		A			C

Second iteration:

Cluster	{(0,0), (2,0)}	{(5,0)}	{(0,4)}	{(4,4)}
Min. Distance	5 to {(5,0)}	5, no update	4 to {(4,4)}, update	4, no update

Update Clusters

Cluster	$\{(0,0), (2,0)\}$	$\{(5,0)\}$	$\{(0,4), (4,4)\}$
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Depiction

D				D	
A		A			C

Third iteration:

Cluster	$\{(0,0), (2,0)\}$	$\{(5,0)\}$	$\{(0,4), (4,4)\}$
Min. Distance	5 to $\{(5,0)\}$	5, no update	8, no update

Update Clusters

Cluster	$\{(0,0), (2,0), (5,0)\}$	$\{(0,4), (4,4)\}$
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Depiction

D				D	
A		A			A

Fourth iteration:

Only two clusters remain, therefore they will be the closest pair.

Update Clusters

Cluster	$\{(0,0), (2,0), (5,0), (0,4), (4,4)\}$
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Since there is only one cluster, the HAC algorithm has completed.

Part B:

Repeat part (a) with the constraint that (2,0) and (5,0) must never be in the same cluster.

Initially, all points are in their own separate clusters.

Depiction of initial state:

D				E	
A		B			C

First iteration:

Cluster	$\{(0,0)\}$	$\{(2,0)\}$	$\{(5,0)\}$	$\{(0,4)\}$	$\{(4,4)\}$
Min. Distance	2 to $\{(2,0)\}$	2, no update	3, no update	4, no update	4, no update

Update Clusters

Cluster	$\{(0,0), (2,0)\}$	$\{(5,0)\}$	$\{(0,4)\}$	$\{(4,4)\}$
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Depiction

D				E	
A		A			C

Second iteration:

Cluster	$\{(0,0), (2,0)\}$	$\{(5,0)\}$	$\{(0,4)\}$	$\{(4,4)\}$
Min. Distance	5 to $\{(5,0)\}$ – ignored 6 to $\{(0,4)\}$	5 to $\{(0,0), (2,0)\}$ – ignored 5 to $\{(4,4)\}$, update	4 to $\{(4,4)\}$, update	4, no update

Update Clusters

Cluster	$\{(0,0), (2,0)\}$	$\{(5,0)\}$	$\{(0,4), (4,4)\}$
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Depiction

D				D	
A		A			C

Third iteration:

Cluster	$\{(0,0), (2,0)\}$	$\{(5,0)\}$	$\{(0,4), (4,4)\}$
Min. Distance	5 to $\{(5,0)\}$ – ignored 8 to $\{(0,4), (4,4)\}$	5 to $\{(0,0), (2,0)\}$ – ignored 9 to $\{(0,4), (4,4)\}$, no update	8, no update

Update Clusters

Cluster	$\{(0,0), (2,0), (0,4), (4,4)\}$	$\{(5,0)\}$
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Depiction

D				D	
A		A			A

All allowed clusters have been formed. Due to the constraint, this iteration is as far as the HAC is allowed to proceed.

Question 3: kNN Classification

Given a dataset with binary labels $(x, y) = \{(2, +), (3, +), (5, -), (7, +), (11, -)\}$, compute the kNN training set error using 0-1 loss and Euclidean distance. If there is a tie, always favor the positive class.

Part A: k=5

- $(2, +)$
- $(3, +)$
- $(5, +)$ Fail
- $(7, +)$
- $(11, +)$ Fail
- Fail: $\frac{2}{5}$

Answer:

$$\text{Error Rate} = \frac{2}{5} = 0.4$$

Part B: k=4

- $(2, +)$
- $(3, +)$
- $(5, +)$ Fail

- (7, +)
- (11, +) Fail
- Fail: $\frac{2}{5}$

Answer:

$$\text{Error Rate} = \frac{2}{5} = 0.4$$

Part C: k=3

- (2, +)
- (3, +)
- (5, +) Fail
- (7, +)
- (11, -) Fail
- Fail: $\frac{1}{5}$

Answer:

$$\text{Error Rate} = \frac{1}{5} = 0.2$$

Part D: k=2

- (2, +)
- (3, +)
- (5, +) Fail
- (7, +)
- (11, +) Fail
- Fail: $\frac{2}{5}$

Answer:

$$\text{Error Rate} = \frac{2}{5} = 0.4$$

Part E: k=1

- (2, +)
- (3, +)
- (5, -)
- (7, +)
- (11, -)
- Fail: $\frac{0}{5}$

Answer:

$$\text{Error Rate} = 0$$

Part F: Should we choose the k with the smallest training set error? Why?

No, we should not choose the k with the smallest training set error. Here, we see that $k=1$ leads to the smallest training set error. This result will be consistent for every training set error calculated with $k=1$ nearest neighbors simply because the only “neighbor” calculated is the point itself within the training set. In a way, we could define the training set error to be zero for $k=1$. So, using $k=1$ for future test instances is by no means guaranteed to be as accurate as the training set error has led us to believe.