

RANKING

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CS 451 – Fall 2013

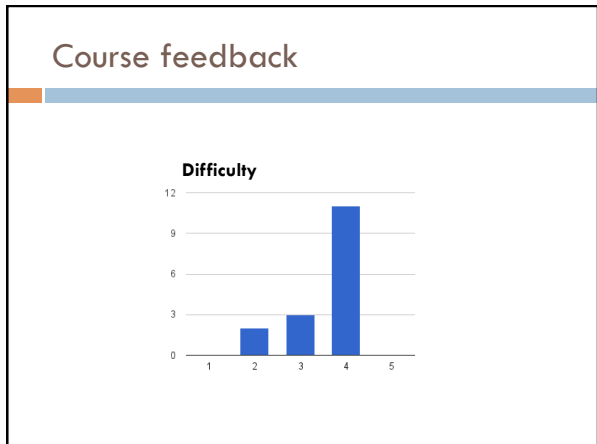
Admin

Assignment 4

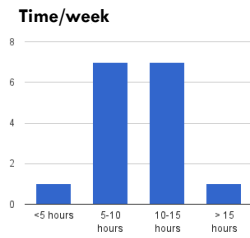
Assignment 5

Course feedback

Thanks!



Course feedback



Course feedback

Experiments/follow-up questions

More interesting data sets

More advanced topics I'm hoping to get to:

- large margin classifiers
- probabilistic modeling
- Unsupervised learning/clustering
- Ensemble learning
- Collaborative filtering
- MapReduce/Hadoop

Java tip for the day

static

```
ArrayList<Example> data = dataset.getData();
```

How can I iterate over it?

Java tip for the day




```
ArrayList<Example> data = dataset.getData();
```

```
for( int i = 0; i < data.size(); i++ ){
    Example ex = data.get(i)
}
```




OR

```
// can do on anything that implements the Iterable interface
for( Example ex: data ){
}
```




An aside: text classification

Raw data	labels
	Chardonnay
	Pinot Grigio
	Zinfandel

Text: raw data




Raw data	labels	Features?
	Chardonnay	
	Pinot Grigio	
	Zinfandel	

Feature examples

Raw data	labels	Features
	Chardonnay	Clinton said pinot repeatedly last week on tv, "pinot, pinot, pinot"
	Pinot Grigio	(1, 1, 1, 0, 0, 1, 0, 0, ...)
	Zinfandel	Occurrence of words

pinot *clinton* *said* *california* *across* *tv* *wrong* *capital*

Feature examples

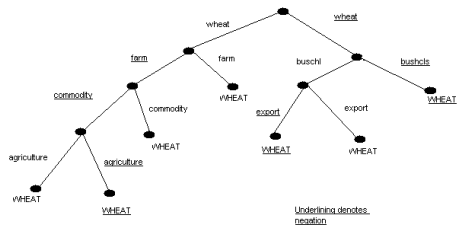
Raw data	labels	Features
	Chardonnay	Clinton said pinot repeatedly last week on tv, "pinot, pinot, pinot"
	Pinot Grigio	(4, 1, 1, 0, 0, 1, 0, 0, ...)
	Zinfandel	Frequency of word occurrences

pinot *clinton* *said* *california* *across* *tv* *wrong* *capital*

This is the representation we're using for assignment 5

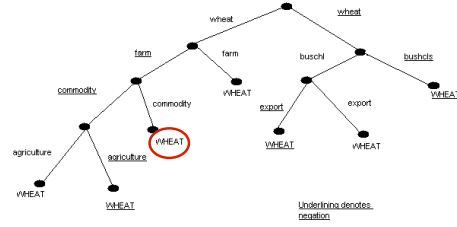
Decision trees for text

Each internal node represents whether or not the text has a particular word



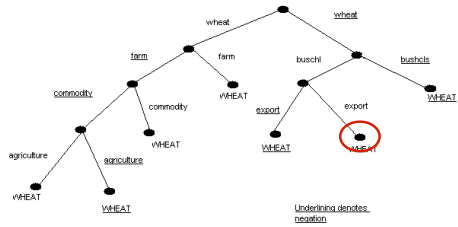
Decision trees for text

wheat is a commodity that can be found in states across the nation

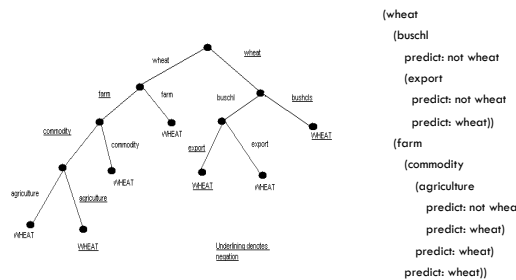


Decision trees for text

The US views technology as a commodity that it can export by the buschl.



Printing out decision trees



Ranking problems

Suggest a simpler word for the word below:

vital

Suggest a simpler word

Suggest a simpler word for the word below:

vital

word	frequency
important	13
necessary	12
essential	11
needed	8
critical	3
crucial	2
mandatory	1
required	1
vital	1

Suggest a simpler word

Suggest a simpler word for the word below:

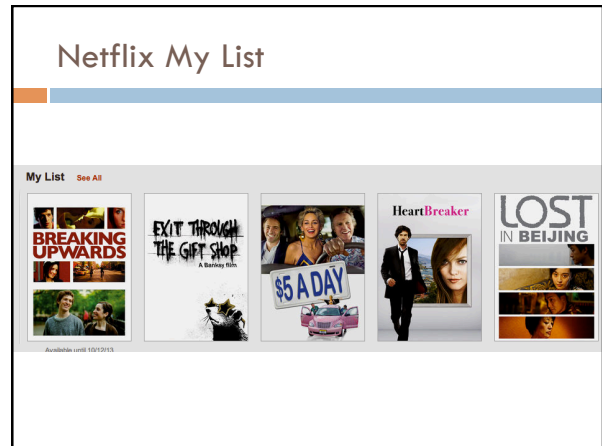
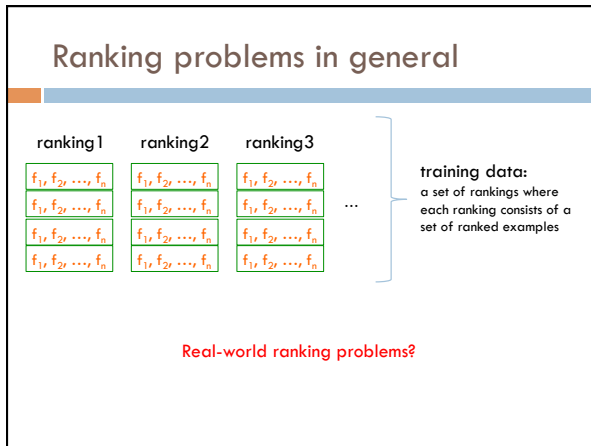
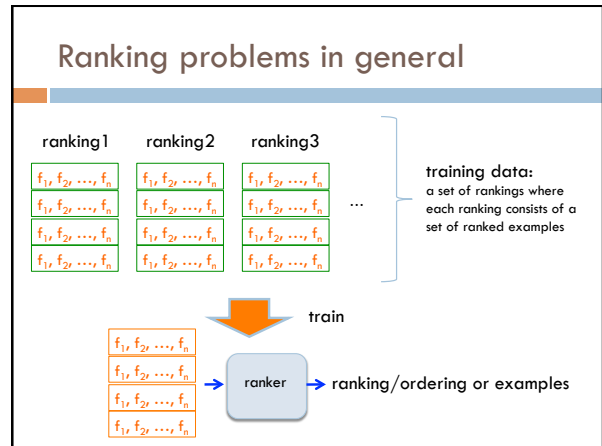
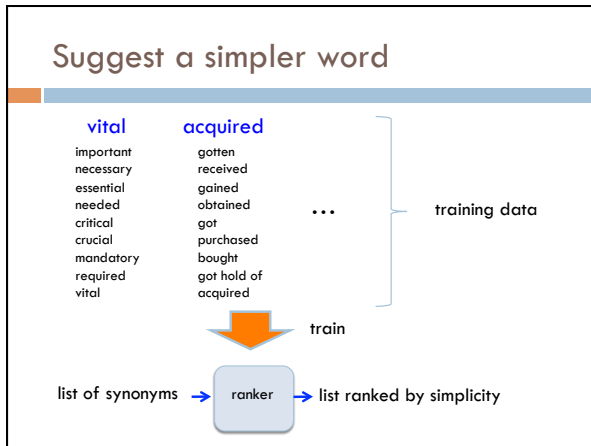
acquired

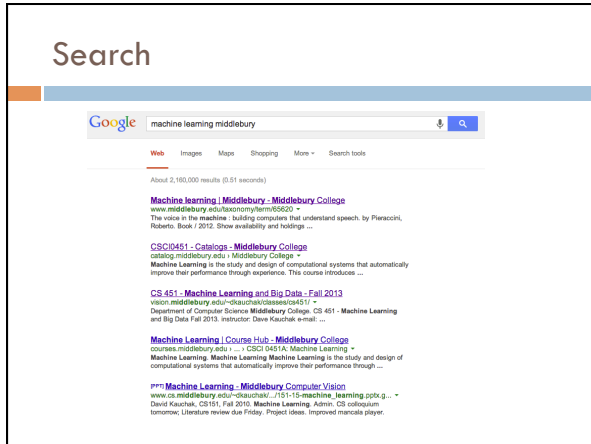
Suggest a simpler word

Suggest a simpler word for the word below:

acquired

word	frequency
gotten	12
received	9
gained	8
obtained	5
got	3
purchased	2
bought	2
got hold of	1
acquired	1





Ranking Applications

- reranking N-best output lists
 - machine translation
 - computational biology
 - parsing
 - ...

- flight search
 - ...

Black box approach to ranking

Abstraction: we have a generic binary classifier, how can we use it to solve our new problem

Can we solve our ranking problem with this?

Predict better vs. worse

Train a classifier to decide if the first input is better than second:

- Consider all possible pairings of the examples in a ranking
- Label as positive if the first example is higher ranked, negative otherwise

ranking 1

f_1, f_2, \dots, f_n
f_1, f_2, \dots, f_n
f_1, f_2, \dots, f_n

Predict better vs. worse

Train a classifier to decide if the first input is better than second:
 - Consider all possible pairings of the examples in a ranking
 - Label as positive if the first example is higher ranked, negative otherwise

ranking1	new examples	binary label
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{11}, f_{21}, \dots, f_{n1}$	+1
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1

Predict better vs. worse

Our binary classifier only takes one example as input

Predict better vs. worse

Our binary classifier only takes one example as input

How can we do this?
 We want features that compare the two examples.

Combined feature vector

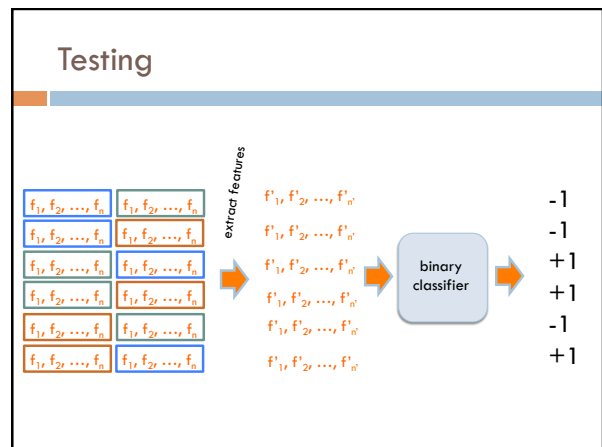
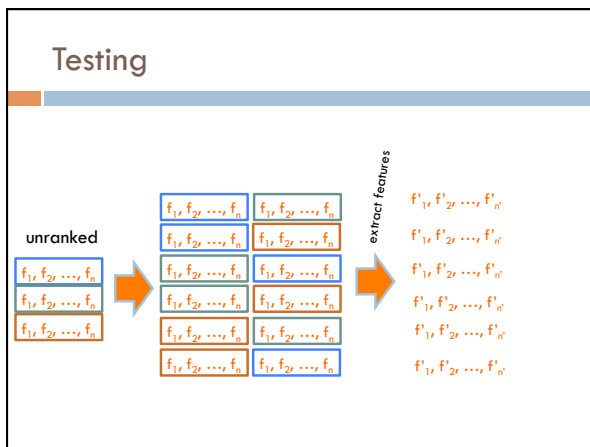
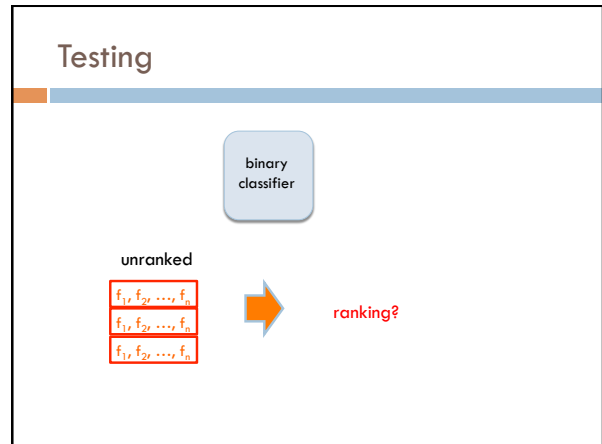
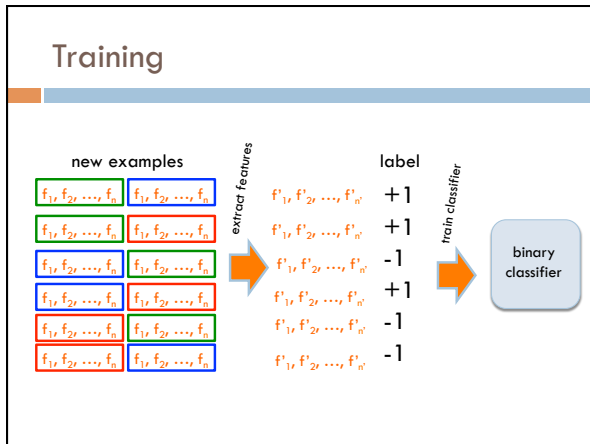
Many approaches! Will depend on domain and classifier

Two common approaches:

1. difference:

$$f'_i = a_i - b_i$$
2. greater than/less than:

$$f'_i = \begin{cases} 1 & \text{if } a_i > b_i \\ 0 & \text{otherwise} \end{cases}$$



Testing

$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1

What is the ranking?
Algorithm?

Testing

for each binary example e_k :
 $label[j] += f_{jk}(e_k)$
 $label[k] -= f_{jk}(e_k)$

rank according to label scores

$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$	+1

→

$f_{11}, f_{21}, \dots, f_{n1}$
$f_{11}, f_{21}, \dots, f_{n1}$
$f_{11}, f_{21}, \dots, f_{n1}$

An improvement?

ranking1	new examples	binary label
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1

Are these two examples the same?

Weighted binary classification

ranking1	new examples	weighted label
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	+2
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	+1
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	-2
$f_{11}, f_{21}, \dots, f_{n1}$	$f_{11}, f_{21}, \dots, f_{n1}$	-1

Weight based on distance in ranking

Weighted binary classification

ranking 1

$f_{11}, f_{21}, \dots, f_{n1}$
$f_{12}, f_{22}, \dots, f_{n2}$
$f_{13}, f_{23}, \dots, f_{n3}$

➔

new examples

$f_{11}, f_{21}, \dots, f_{n1}$	$f_{12}, f_{22}, \dots, f_{n2}$
$f_{12}, f_{22}, \dots, f_{n2}$	$f_{13}, f_{23}, \dots, f_{n3}$
$f_{13}, f_{23}, \dots, f_{n3}$	$f_{14}, f_{24}, \dots, f_{n4}$
$f_{14}, f_{24}, \dots, f_{n4}$	$f_{15}, f_{25}, \dots, f_{n5}$
$f_{15}, f_{25}, \dots, f_{n5}$	$f_{16}, f_{26}, \dots, f_{n6}$

weighted label

+1
+2
-1
+1
-2
-1

In general can weight with any consistent distance metric

Can we solve this problem?

Testing

If the classifier outputs a confidence, then we've learned a *distance* measure between examples

During testing we want to rank the examples based on the learned distance measure

Ideas?

Testing

If the classifier outputs a confidence, then we've learned a *distance* measure between examples

During testing we want to rank the examples based on the learned distance measure

Sort the examples and use the output of the binary classifier as the similarity between examples!

Ranking evaluation

	ranking	prediction
$f_{11}, f_{21}, \dots, f_{n1}$	1	1
$f_{12}, f_{22}, \dots, f_{n2}$	2	3
$f_{13}, f_{23}, \dots, f_{n3}$	3	2
$f_{14}, f_{24}, \dots, f_{n4}$	4	5
$f_{15}, f_{25}, \dots, f_{n5}$	5	4

Ideas?

Idea 1: accuracy

	ranking	prediction	
$f_{1r}, f_{2r}, \dots, f_{nr}$	1	1	$1/5 = 0.2$
$f_{1r}, f_{2r}, \dots, f_{nr}$	2	3	
$f_{1r}, f_{2r}, \dots, f_{nr}$	3	2	
$f_{1r}, f_{2r}, \dots, f_{nr}$	4	5	
$f_{1r}, f_{2r}, \dots, f_{nr}$	5	4	

Any problems with this?

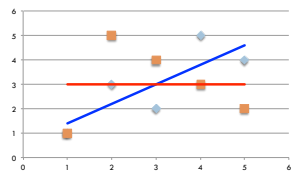
Doesn't capture "near" correct

	ranking	prediction	prediction
$f_{1r}, f_{2r}, \dots, f_{nr}$	1	1	1
$f_{1r}, f_{2r}, \dots, f_{nr}$	2	3	5
$f_{1r}, f_{2r}, \dots, f_{nr}$	3	2	4
$f_{1r}, f_{2r}, \dots, f_{nr}$	4	5	3
$f_{1r}, f_{2r}, \dots, f_{nr}$	5	4	2

$1/5 = 0.2$

Idea 2: correlation

ranking	prediction	prediction
1	1	1
2	3	5
3	2	4
4	5	3
5	4	2



Look at the correlation between the ranking and the prediction