

ColumbiaX: CSMM.101x Artificial Intelligence (AI)

Help

Week 6: Machine Learning 2 > Week 6 Quiz: Machine Learning 2 > Week 6 Quiz Bookmarks Week 6 Quiz ☐ Bookmark this page Artificial Intelligence True/False Course: 16.3/16.3 points (graded) Getting When a decision tree is fully grown, it is likely to overfit the training data. Started False ▶ Week 1: Introduction to ΑI True ▶ Week 2: Intelligent Agents and Uninformed Submit You have used 1 of 1 attempt Search ▶ Week 3: True/False Heuristic 10.0/10.0 points (graded) Search Logistic regression is a linear classification method. ▶ Week 4: False Adversarial Search and True Games ▶ Week 5: Machine Learning 1 Submit You have used 1 of 1 attempt ▼ Week 6: Machine Multiple choice Learning 2 10.0/10.0 points (graded) The idea of boosting is to train weak learners on weighted training Week 6: examples. Check all that apply. Suggested Readings 6.1 Machine Give large weights to easy examples to get rid of them Learning: Perceptron

Week 6 Quiz | Week 6 Quiz: Machine Learning 2 | CSMM.101x Courseware | edX

- 6.2 Logistic Regression
- 6.3 Decision Trees
- 6.4 Naive Bayes
- 6.5 Ensemble Methods

Week 6 Quiz: Machine Learning 2 Quiz due Apr 11, 2017

Week 6 Discussion Questions

05:00 IST

 Use any classifier as far as its accuracy is slightly worse than random

✓ The classification output is a majority voting of all weak classifiers outputs ✓

☑ Give large weights to hard examples to focus on those in the next steps ✓

V

Submit

You have used 1 of 1 attempt

True/False

10.0/10.0 points (graded)

A decision tree generated from unbalanced training data may be biased towards the majority class.

False

True

Submit

You have used 1 of 1 attempt

Checkboxes

10.0/10.0 points (graded) Check all that apply.

- Naive Bayes classifier models p(y|x) and p(y) and then used Bayes rule to obtain p(x|y)
- Naive Bayes classifier models p(x|y) and p(y) and then used Bayes rule to obtain p(y|x)

Naive Bayes classifier models $p(x y)$ and $p(x)$ and then used
Bayes rule to obtain p(x y)

- Naive Bayes classifier is a discriminative method.
- Naive Bayes classifier is a generative method.



Submit

You have used 1 of 1 attempt

True/False

10.0/10.0 points (graded)

a	$\mid b \mid$	f
1	1	1
1	0	0
0	1	1
0	0	0

Can the boolean function f be represented with a perceptron?



False

Submit

You have used 1 of 1 attempt

Checkboxes

0.0/10.0 points (graded)

Consider the toy example in slide 8, decision trees handout. Suppose there is one additional feature "application_number". Suppose all the examples in this dataset have different values {v1, v2, v3, ..., v14} for application_number, corresponding to the order they appear in the table.

- application_number will have the lowest gain and hence will never be picked at the root
- application_number will be picked at the root as a first choice to split the data because it has the highest gain
- application_number is the most discriminative feature
- application_number is a useless feature and should be discarded



Submit

You have used 1 of 1 attempt

Checkboxes

10.0/10.0 points (graded)

Consider the toy example in slide 8, decision trees handout. Suppose there is one additional feature "application_number". Suppose all the examples in this dataset have different values {v1, v2, v3, ..., v14} for application_number, corresponding to the order they appear in the table.

Now, we define a new feature called "even_or_odd", which takes on the value "even" for {v2, v4, v6, ..., v14}, and takes on the value "odd" for {v1, v3, v5, ..., v13}.

- $\ lue{}$ even_or_odd will have the same gain as "Highest Degree"
- $\hfill \square$ even_or_odd will have the same gain as "Work Experience"
- lacktriangledown even_or_odd will have the same gain as "Favorite Language" lacktriangledown
- even_or_odd will have the same gain as "Needs Work Visa"

