FUNDAMENTALS OF MACHINE LEARNING

SCIKIT-LEARN ADDITIONAL LECTURE

Metrics

- How to evaluate how good or effective is your model?
- Various performance metrics:
 - Confusion matrix
 - Accuracy vs. precision vs. recall
 - F1 score

Example: Confusion matrix

- □ It is a multi-class classification problem Example of 3 classes of professors
- An example of a confusion matrix C where $C_{ij} = \# of \ observations \ of \ category \ j \ but \ the \ true \ category \ is \ i$

Predict / Actual	Human	Octopus	Penguin
Human	5	0	2
Octopus	3	3	3
Penguin	0	1	11

Good for penguin, not so good in predicting octopus

Predict / Actual		Positive	Negative	
Positive	C	1	2	
Negative		0	997	

- Accuracy: correct prediction / all observations
- □ In this case, the accuracy is (1+997)/(1+2+0+997)=.998
- But two thirds of the people who actually have the disease will be judged free of it by this model !!!!

Predict / Actual	Positive	Negative
Positive	1	2
Negative	0	997

Positive	C ₀₀	C ₀₁
Negative	C ₁₀	C ₁₁

- □ Precision (of positive) = true positive /(true positive + false positive)
 What is the precision of "negative"?
- □ For the above case, it is 1, best we can have ☺
- $\Box \operatorname{Precision}_{i} = \frac{c_{ii}}{\sum_{j} c_{ji}} \quad \text{for any class } i$

	Predict / Actual	Positive	Negative
С	Positive	1	2
	Negative	0	997

Positive	C ₀₀	C ₀₁
Negative	C ₁₀	C ₁₁

- Our problem is not in precision, but in the recall
- Recall (of positive) = true positive /(true positive + false negative) = 1/(1+2) = 1/3
- □ For the above case, our recall on positive is only 1/3 accurate !!!
- $\square \operatorname{Recall}_{i} = \frac{c_{ii}}{\sum_{i} c_{ij}} \quad \text{for any class } i$

Predict / Actual	Positive	Negative
Positive	1	2
Negative	0	997

- □ F1 score: combine precision and recall to produce a single score
 Why F1 score is intuitive?
- □ $F1 = 2 * \frac{precision * recall}{precision + recall}$ for each category
- extstyle extstyle extstyle extstyle F1 scores for all categories