

For a more in-depth introduction to Naïve Bayes Classifiers and the theory surrounding them, please see Andrew's lecture on Probability for Data Miners.

Naïve Bayes Classifiers

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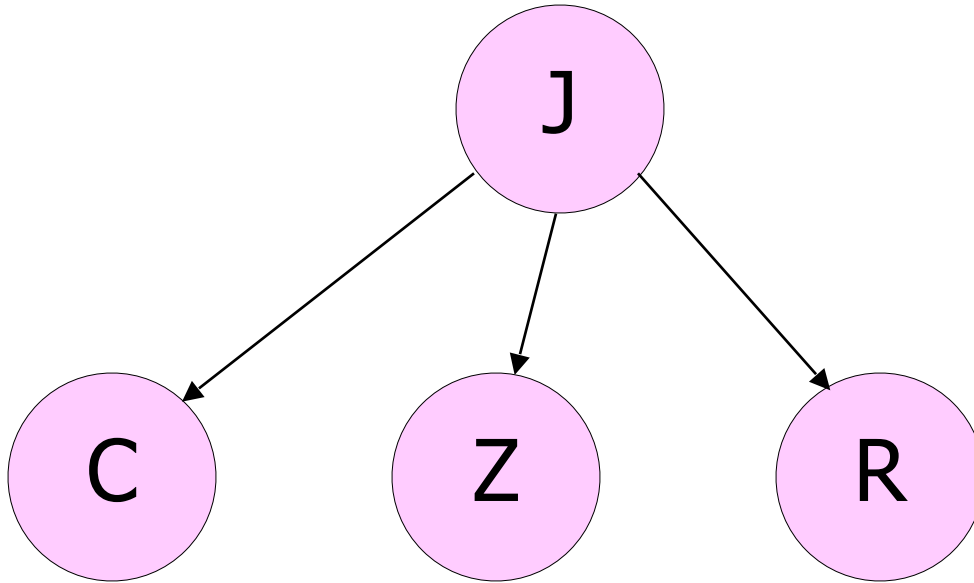
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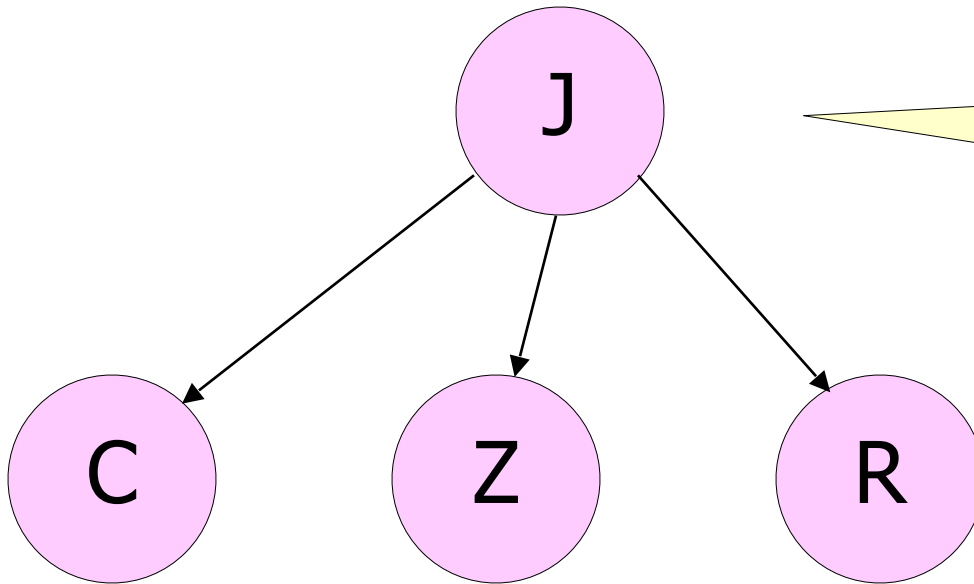
These notes assume you have already met Bayesian Networks

A simple Bayes Net



J	Person is a Junior
C	Brought Coat to Classroom
Z	Live in zipcode 15213
R	Saw "Return of the King" more than once

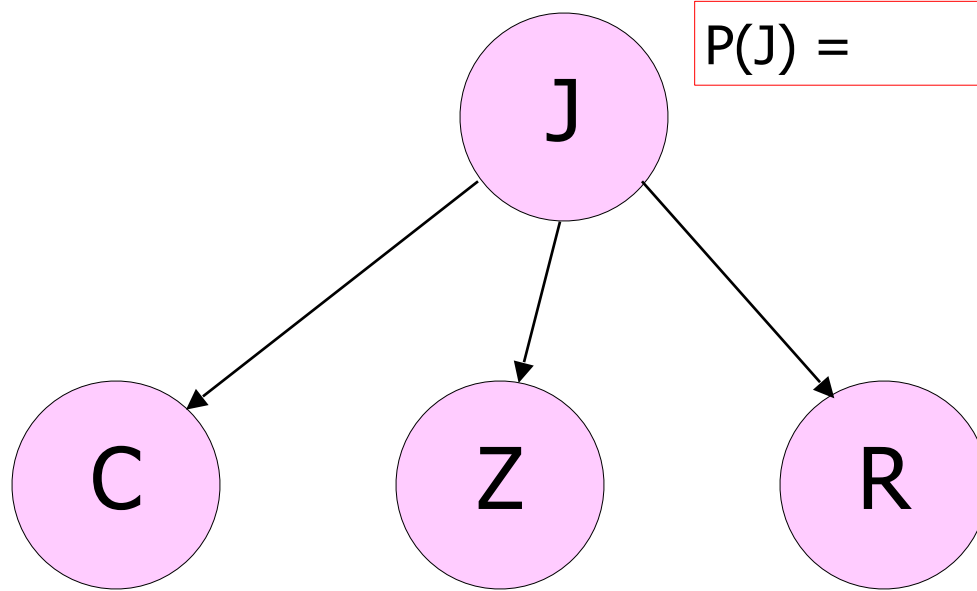
A simple Bayes Net



What parameters are stored in the CPTs of this Bayes Net?

J	Person is a Junior
C	Brought Coat to Classroom
Z	Live in zipcode 15213
R	Saw "Return of the King" more than once

A simple Bayes Net



$$P(J) =$$

J	Person is a Junior
C	Brought Coat to Classroom
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$$P(C|J) =$$

$$P(C|\sim J) =$$

$$P(Z|J) =$$

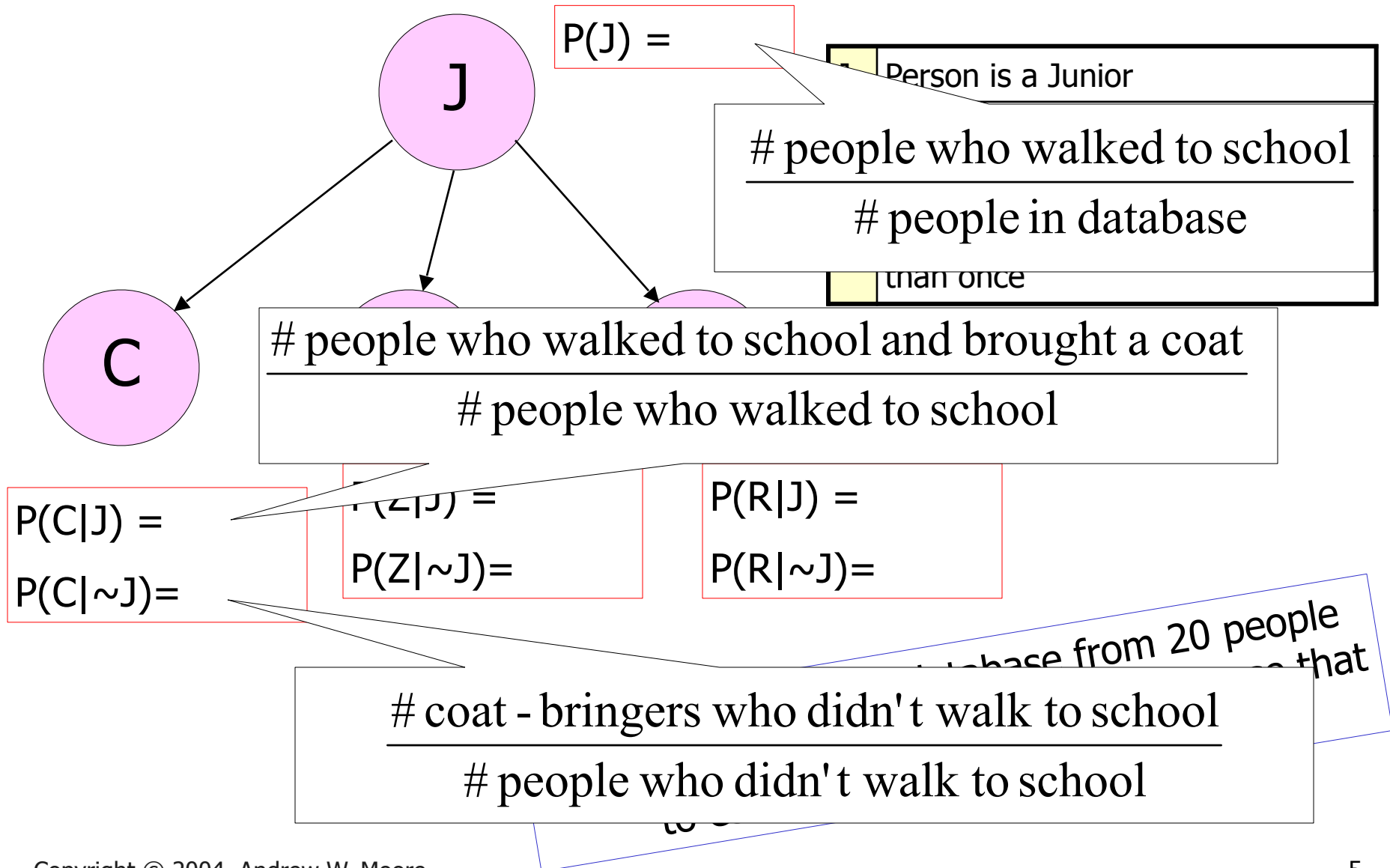
$$P(Z|\sim J) =$$

$$P(R|J) =$$

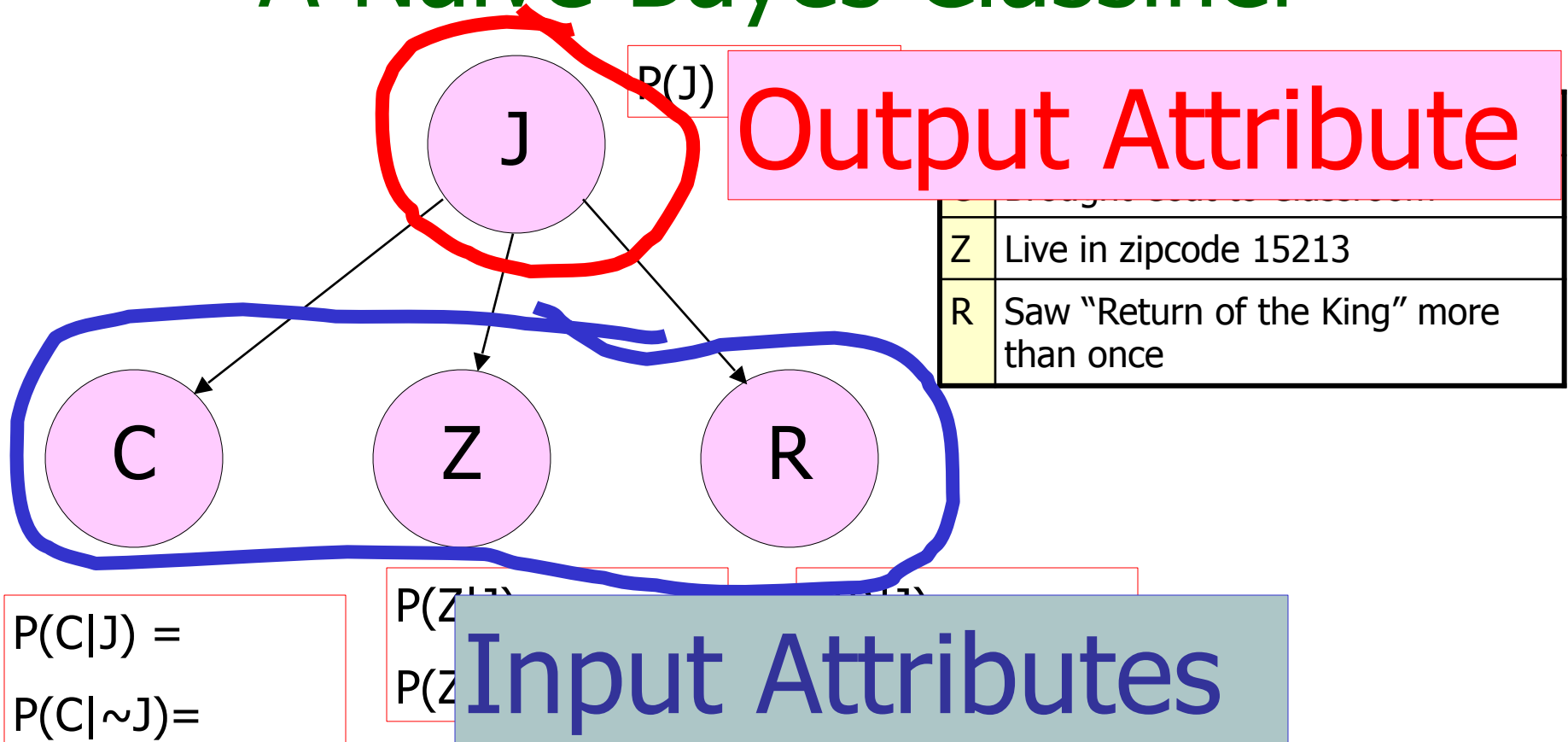
$$P(R|\sim J) =$$

Suppose we have a database from 20 people who attended a lecture. How could we use that to estimate the values in this CPT?

A simple Bayes Net



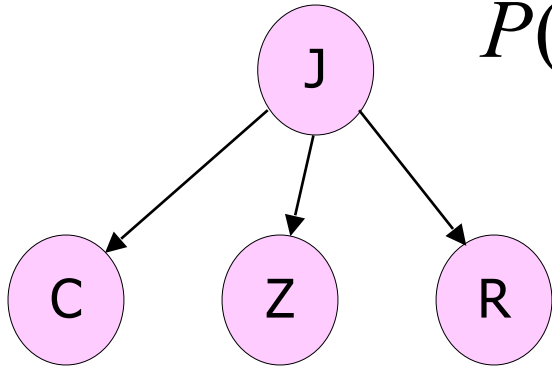
A Naïve Bayes Classifier



A new person shows up at class wearing an "I live right above the Manor Theater where I saw all the Lord of The Rings Movies every night" overcoat.

What is the probability that they are a Junior?

Naïve Bayes Classifier Inference



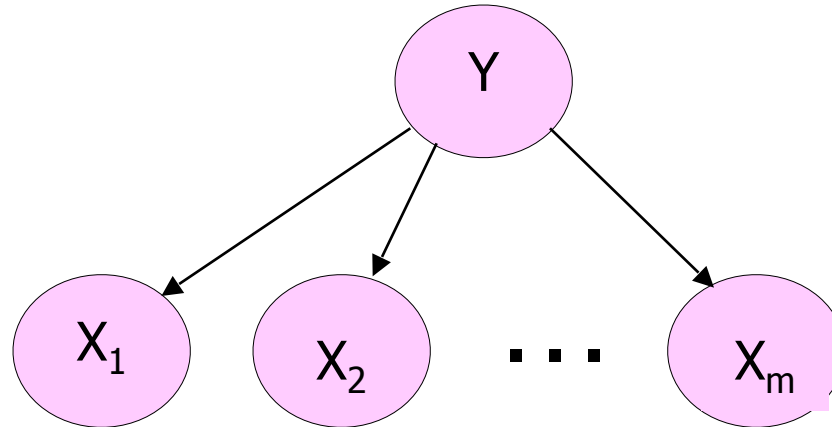
$$P(J \mid C \wedge \neg Z \wedge R) =$$

$$= \frac{P(J \wedge C \wedge \neg Z \wedge R)}{P(C \wedge \neg Z \wedge R)}$$

$$= \frac{P(J \wedge C \wedge \neg Z \wedge R)}{P(J \wedge C \wedge \neg Z \wedge R) + P(\neg J \wedge C \wedge \neg Z \wedge R)}$$

$$= \frac{P(C \mid J)P(\neg Z \mid J)P(R \mid J)P(J)}{\left(\begin{array}{c} P(C \mid J)P(\neg Z \mid J)P(R \mid J)P(J) \\ + \\ P(C \mid \neg J)P(\neg Z \mid \neg J)P(R \mid \neg J)P(\neg J) \end{array} \right)}$$

The General Case



1. Estimate $P(Y=v)$ as fraction of records with $Y=v$
2. Estimate $P(X_i=u \mid Y=v)$ as fraction of “ $Y=v$ ” records that also have $X=u$.
3. To predict the Y value given observations of all the X_i values, compute

$$Y^{\text{predict}} = \underset{v}{\operatorname{argmax}} P(Y = v \mid X_1 = u_1 \boxed{?} X_m = u_m)$$

Naïve Bayes Classifier

$$Y^{\text{predict}} = \underset{v}{\operatorname{argmax}} P(Y = v \mid X_1 = u_1 \boxed{?} X_m = u_m)$$

$$Y^{\text{predict}} = \underset{v}{\operatorname{argmax}} \frac{P(Y = v \wedge X_1 = u_1 \boxed{?} X_m = u_m)}{P(X_1 = u_1 \boxed{?} X_m = u_m)}$$

$$Y^{\text{predict}} = \underset{v}{\operatorname{argmax}} \frac{P(X_1 = u_1 \boxed{?} X_m = u_m \mid Y = v) P(Y = v)}{P(X_1 = u_1 \boxed{?} X_m = u_m)}$$

$$Y^{\text{predict}} = \underset{v}{\operatorname{argmax}} P(X_1 = u_1 \boxed{?} X_m = u_m \mid Y = v) P(Y = v)$$

Because of the structure of the Bayes Net

$$Y^{\text{predict}} = \underset{v}{\operatorname{argmax}} P(Y = v) \prod_{j=1}^{n_Y} P(X_j = u_j \mid Y = v)$$

More Facts About Naïve Bayes Classifiers

- Naïve Bayes Classifiers can be built with real-valued inputs*
- Rather Technical Complaint: Bayes Classifiers don't try to be maximally discriminative---they merely try to honestly model what's going on*
- Zero probabilities are painful for Joint and Naïve. A hack (justifiable with the magic words "Dirichlet Prior") can help*.
- Naïve Bayes is wonderfully cheap. And survives 10,000 attributes cheerfully!

*See future Andrew Lectures

What you should know

- How to build a Bayes Classifier
- How to predict with a BC