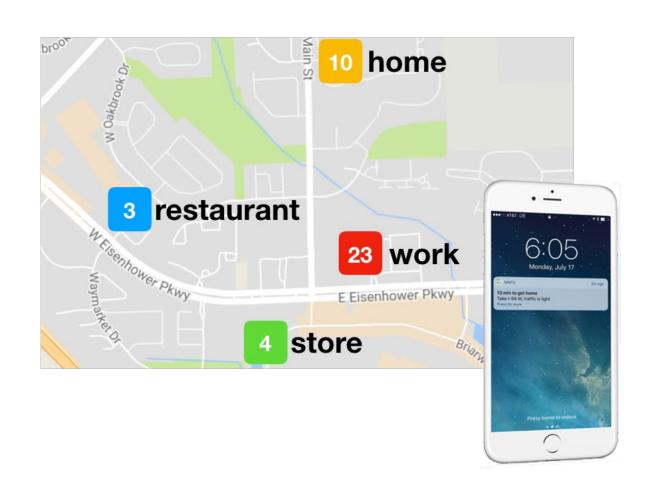




# Understanding Bayesian methods

Brett Lantz Instructor

#### Estimating probability

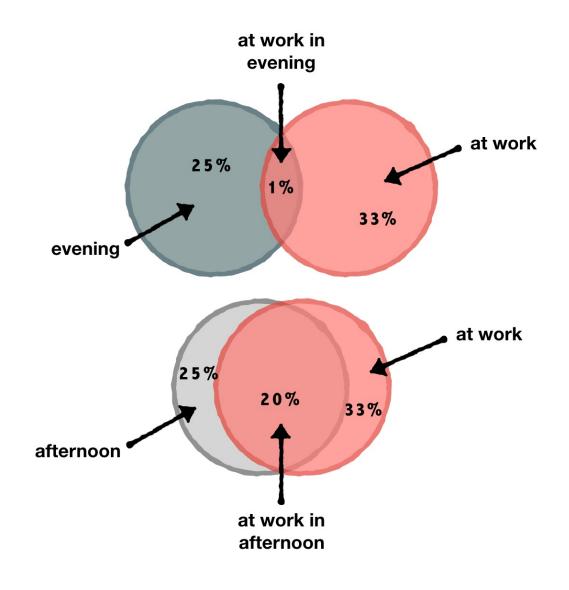


The **probability** of A is denoted P(A)

- P(work) = 23 / 40 = 57.5%
- P(store) = 4 / 40 = 10.0%



### Joint probability and independent events

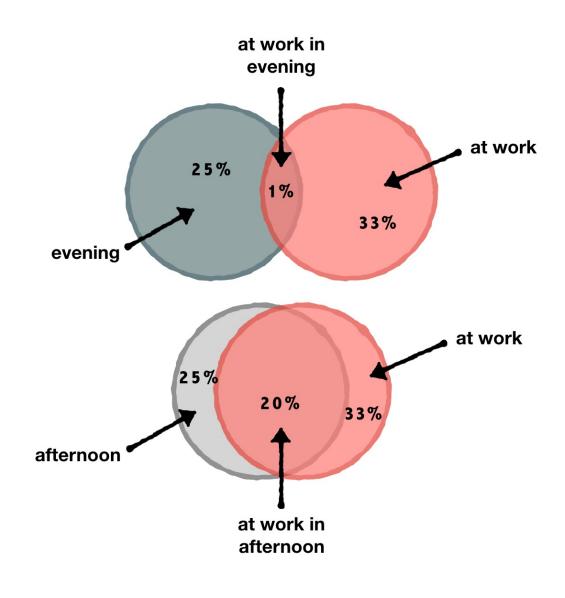


The **joint probability** of events A and B is denoted P(A and B)

- P(work and evening) = 1%
- P(work and afternoon) = 20%



#### Conditional probability and dependent events



The **conditional probability** of events A and B is denoted P(A | B)

- P(A | B) = P(A and B) / P(B)
- P(work | evening) = 1 / 25 = 4%
- P(work | afternoon) = 20 / 25 =80%



#### Making predictions with Naive Bayes

```
# building a Naive Bayes model
library(naivebayes)
m <- naive_bayes(location ~ time_of_day, data = location_history)

# making predictions with Naive Bayes
future_location <- predict(m, future_conditions)</pre>
```





# Let's practice!



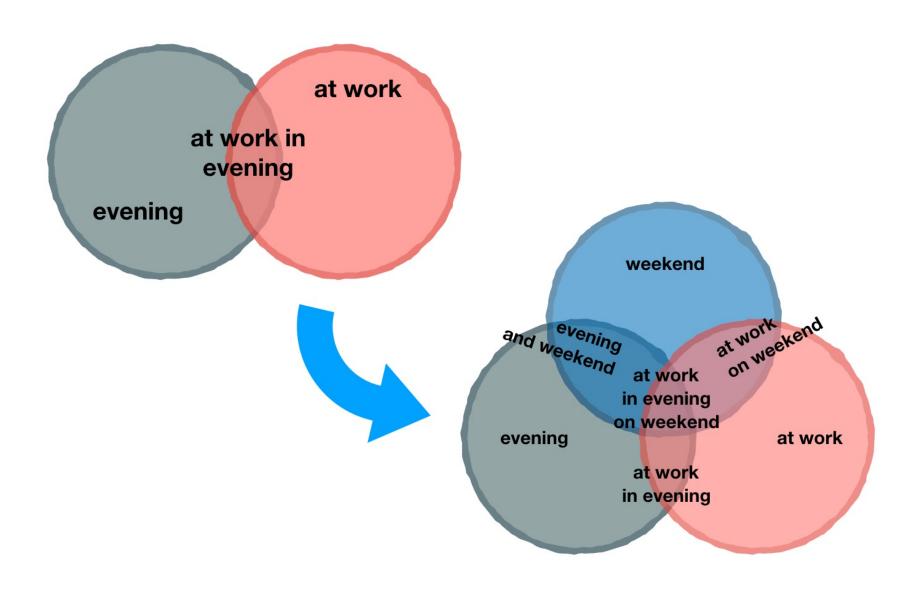


# **Understanding NB's**"naivety"

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Instructor

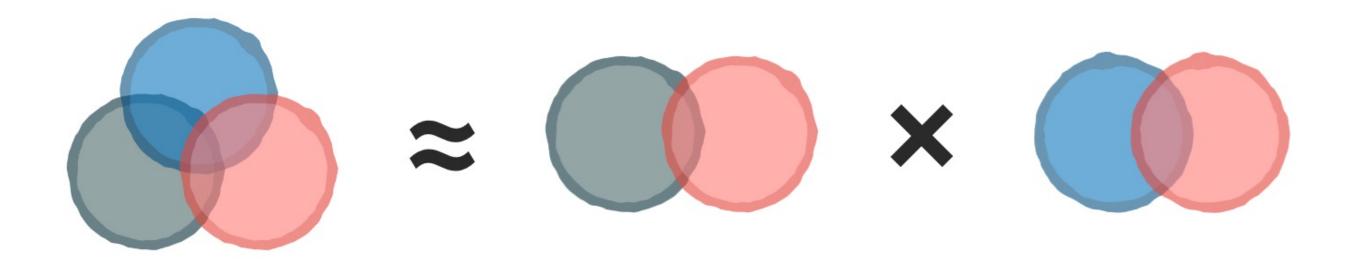


### The challenge of multiple predictors



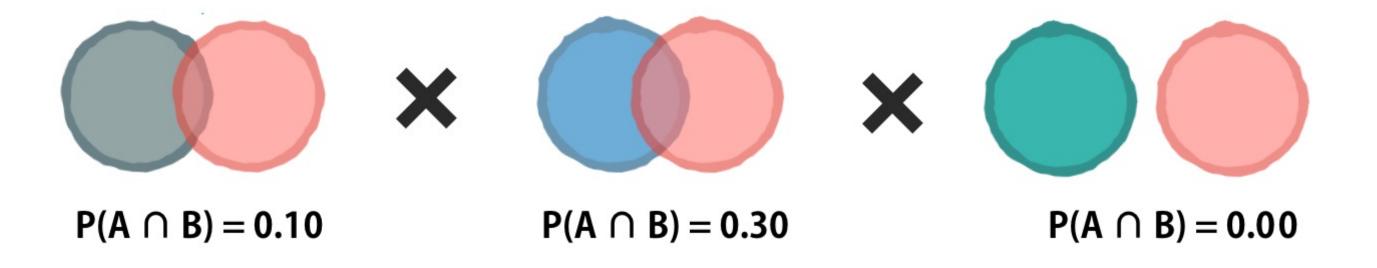


## A "naive" simplification



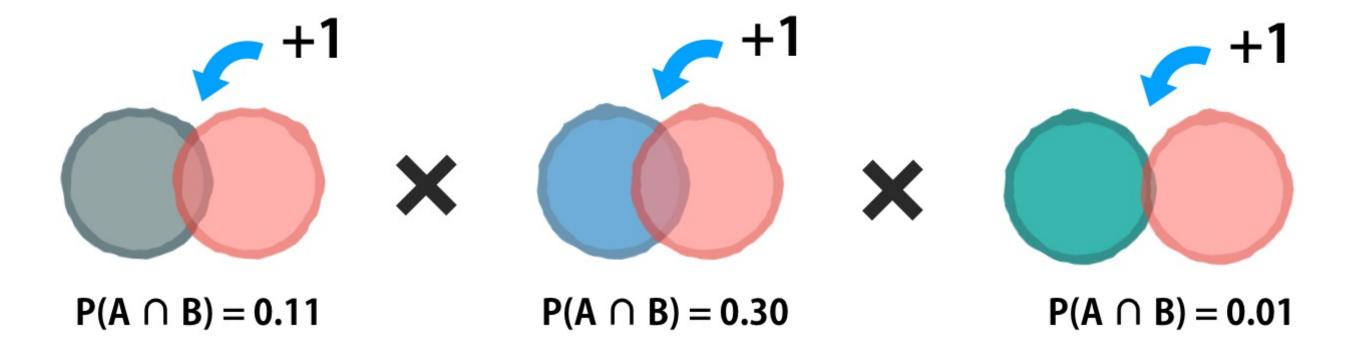


#### An "infrequent" problem





#### The Laplace correction







# Let's practice!



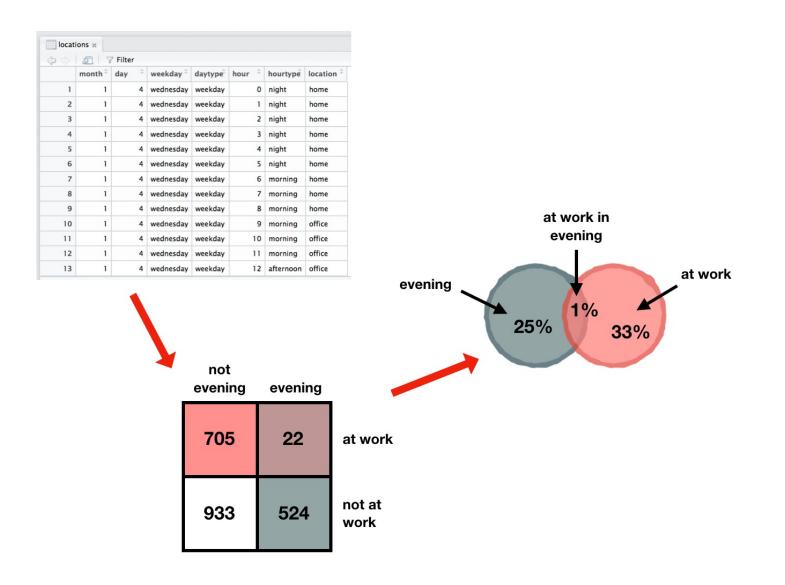


# Applying Naive Bayes to other problems

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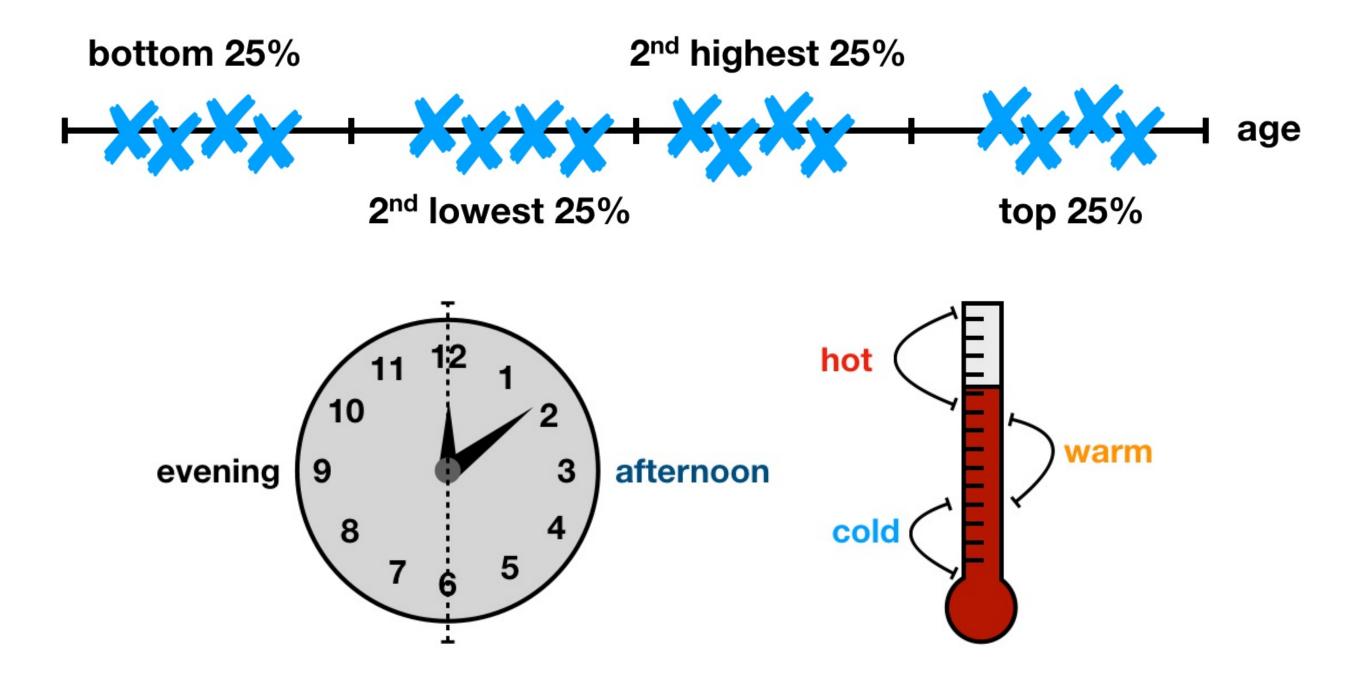


### How Naive Bayes uses data



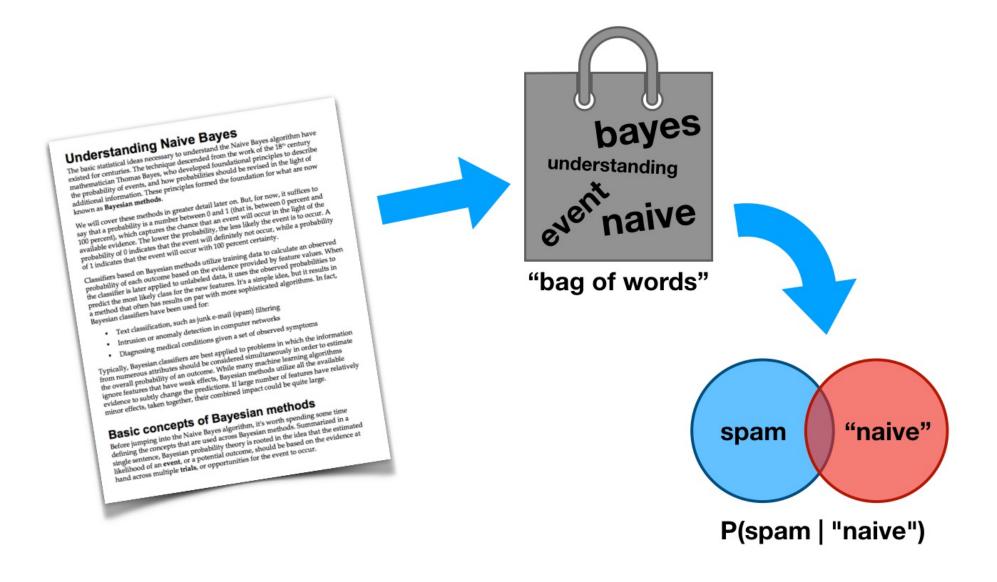


#### Binning numeric data for Naive Bayes





#### Preparing text data for Naive Bayes







# Let's practice!