Understanding Bayesian methods

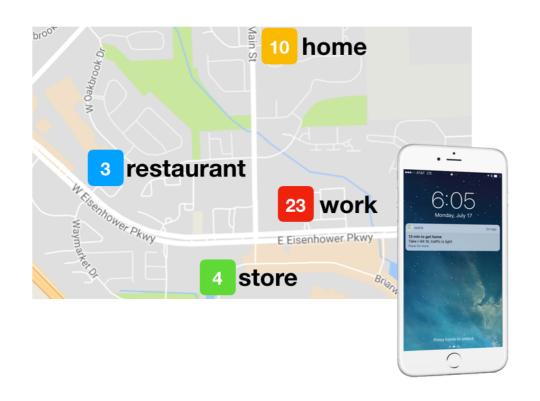
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Brett Lantz
Instructor



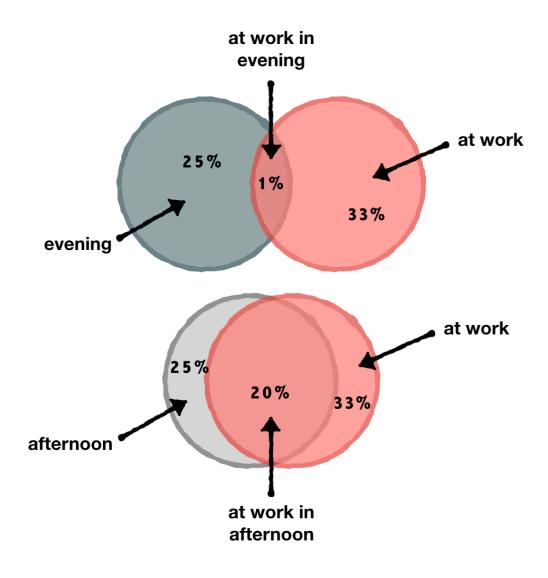
Estimating probability



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

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

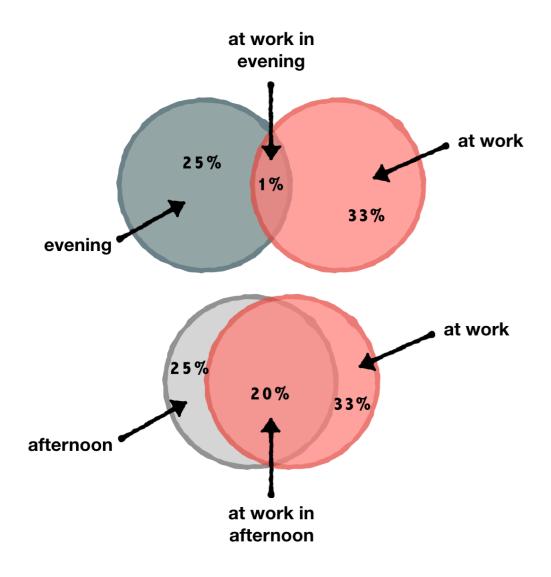
Joint probability and independent events



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 =
- 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)</pre>
```

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

Let's practice!

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Understanding NB's "naivety"

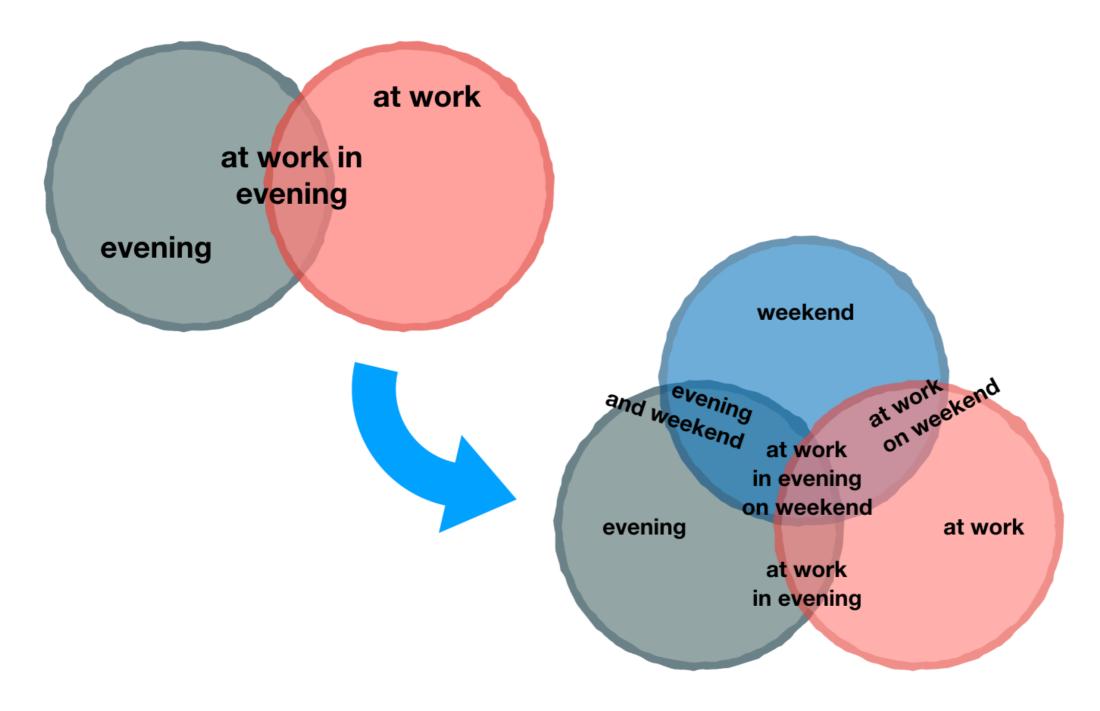
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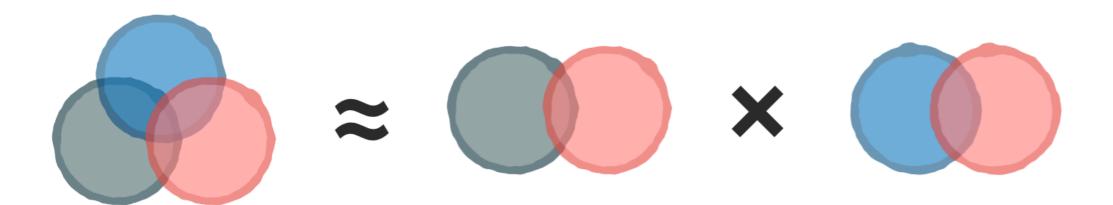


The challenge of multiple predictors

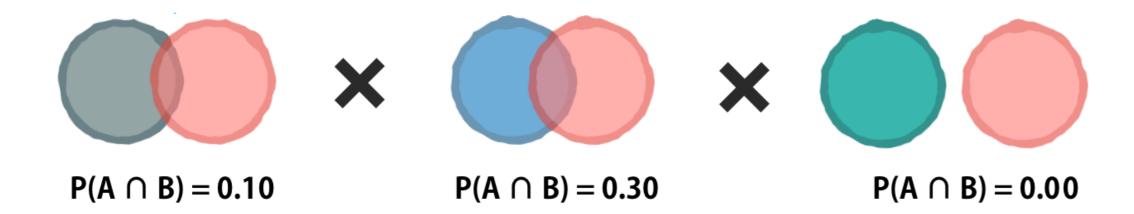




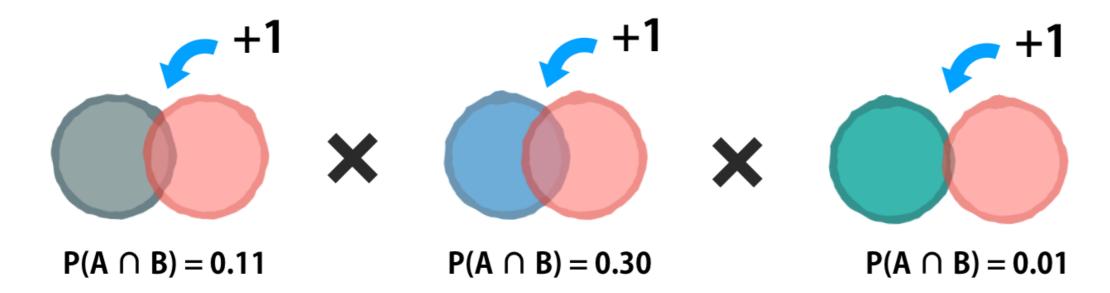
A "naive" simplification



An "infrequent" problem



The Laplace correction



Let's practice!

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Applying Naive Bayes to other problems

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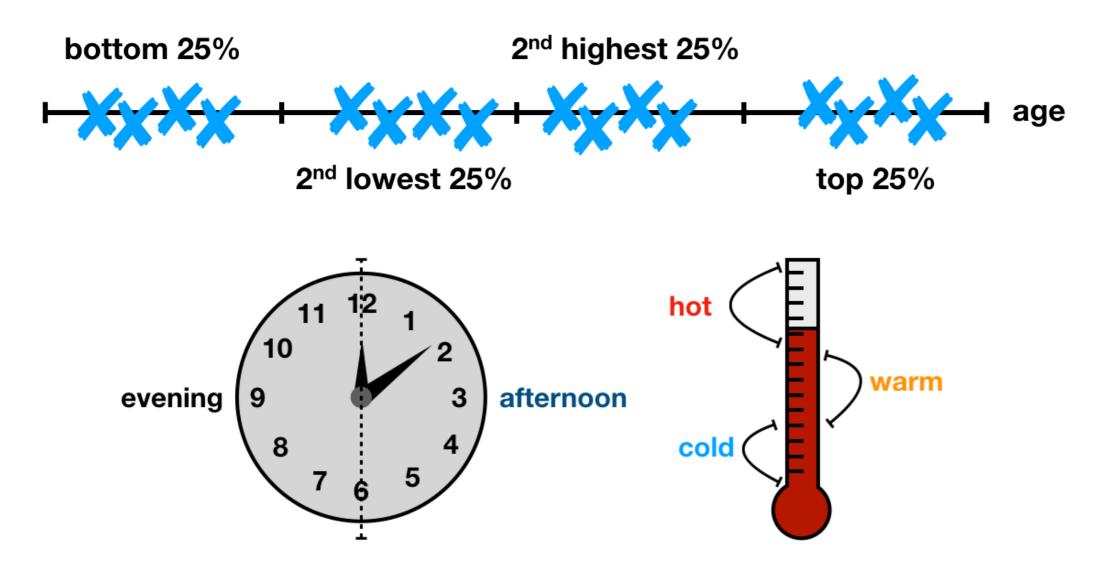
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Instructor



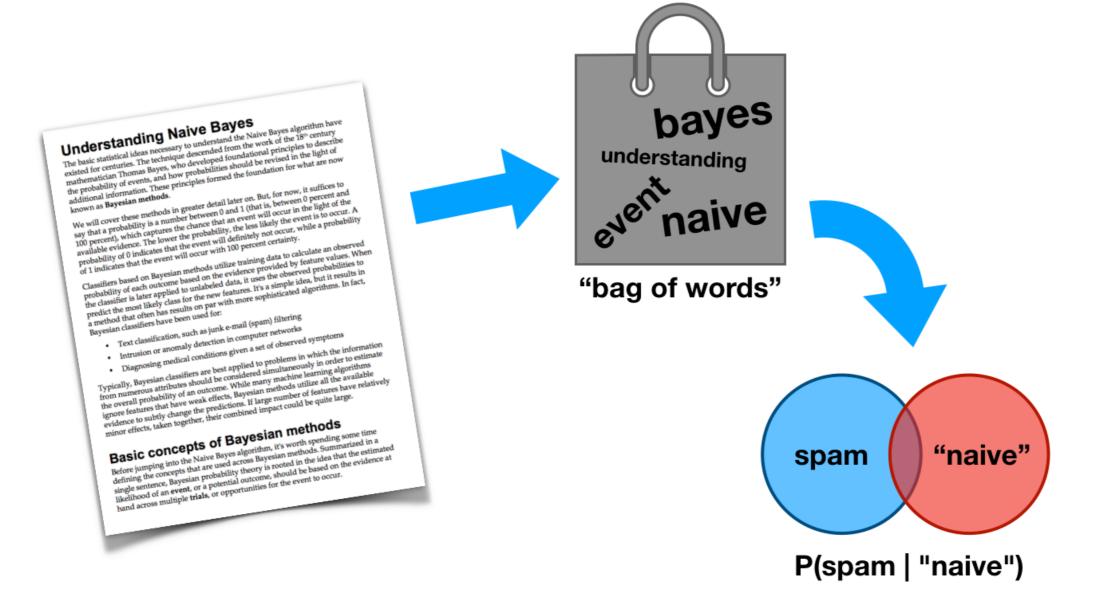
How Naive Bayes uses data



Binning numeric data for Naive Bayes



Preparing text data for Naive Bayes



Let's practice!

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