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Scribing Notes for 3/30

- 1) Today will consist of an exploration of peakdemand data set
- 2) For a while we will close books on regression analysis until later
- 3) Today we start new unit on probability & risk modeling, including new tools and vocabulary to learn
- 4) Professor Scott started with a brief explanation of next homework on environmentally friendly commercial real estate properties; task: recommend action → use regression model and put into context of decision of what should be done; this assignment will incorporate everything we have learned

Import peakdemand.csv

Task: build good predictive model (forecasting) for peakdemand

R Code:

```
summary(peakdemand)
```

```
plot(peakdemand$PeakDemand)
```

```
# Time or period index
```

```
peakdemand$Period = 1:nrow(peakdemand)
```

```
lm1 = lm(PeakDemand ~ Period, data=peakdemand)
```

```
plot(PeakDemand ~ Period, data=peakdemand)
```

```
abline(lm1, col='red', lwd=4)
```

```
lm2 = lm(PeakDemand ~ Period + DailyTemp + I(DailyTemp^2) + Sat +  
Sun, data=peakdemand)
```

```
plot(resid(lm1), type='l')
```

```
lines(resid(lm2), col='red')
```

```
lm3 = lm(PeakDemand ~ Period + DailyTemp + I(DailyTemp^2) + Sat +  
Sun + factor(Month), data=peakdemand)
```

```
plot(resid(lm3), type='l')
```

```
summary(lm3)
```

```
anova(lm3)
```

```
lmstep= step(lm3, direction='backward')
```

```
plot(resid(lm3), type='l')
```

```
# Compare fits
```

```
plot(PeakDemand ~ Period, data=peakdemand,type='l')
lines(fitted(lm3) ~ Period, data=peakdemand, col='red')
```

- 1) Plot peakdemand over time (time index) → notice somewhat upward trend (seasonal effect)
- 2) Hotter months → higher demand
- 3) Adjusting for dependence of peakdemand on temperature, then we won't see seasonal effects right?
- 4) What are some other effects besides upward trend and seasonal? Plot peakdemand vs dailytemp → temperature “smile” → energy usage not as high during the fall
- 5) What do you do with non-linear relationships? (peakdemand vs dailytemp) → add polynomial terms (utilities data set)
- 6) What are some more effects to consider? Sundays energy usage is lower → people are out of house more, big businesses are closed, industries are less heavily engaged on Sundays, etc.
 - a) This suggests micro-cyclical patterns → Must implement dummy variable strategy with Saturdays vs. Sundays
- 7) Now that we are addressing these affects, we must build a model that incorporates all of these effects**
- 8) Let's model trend and seasonal effects
 - a) Trend strategy: Regress time index
 - b) Seasonal strategy: Incorporate dummy variables
- 9) As we transition from lm1, to lm2, and to lm3, we see better predictive models
 - a) Compare fit (red vs black) between “plot(PeakDemand ~ Period, data=peakdemand,type='l')” and “lm3”

New material

- 1) Probability → language of uncertainty; coverage intervals; emphasize data exploration; general concept for quantifying uncertainty
- 2) Must become familiar with terminology
- 3) Basic Rules (Kolmogorov's Axioms)
 - a) Probabilities must sum to 1
 - b) Probabilities for disjoint events (cannot occur at once) add together
 - c) Probability are numbers between 0 and 1 (will sometimes be a percentage)
- 4) More complex rules
 - a) Addition rule (Union Rule)
 - i) $P(A \cup B) = P(A) + P(B) - P(A, B)$ → (Probability that you are a lady and are from Dallas and you subtract $P(A, B)$ because otherwise you double count ladies from Dallas) → “A&B” joint event otherwise probability is zero
 - b) Multiplication rule
 - i) $P(A, B) = P(A) * P(B|A)$ → **conditional upon or given**; probability of A and B = Probability of A times Probability of B **given A** → “probability of getting sued and lose at trial,” **probability of getting sued times probability of getting sued and probability of losing at trial**

- ii) The idea is to build scenarios with preceding sequence of events
- 5) "But what does it all mean??"
 - a) Frequency interpretation of probability → Vegas interpretation
 - i) $P(\text{Black 31}) = \# \text{ Times Black 31 comes up} / \# \text{ spins of roulette wheel}$
 - b) Degree of belief interpretation → formal mathematical argument in course notes if you wish to take a look
 - i) What is chance that it will rain today? 10%?
 - ii) How do we make it slightly more specific? Talk about behavior not brains
 - iii) \$100 contract that it will rain between now and 5; how much are you going to pay me to hold this contract? If it rains, you get \$100, so you pay \$10 to hold contract? Subjective assessment of 10% can become more precise → amount you would pay to hold contract if it comes true; take assessment of your particular belief → revise probability lower if you don't want to pay as much
 - iv) Wall Street interpretation → futures contracts are being bought and sold

Non-trivial rule, but most important rule

- c) Bayes' Rule
 - i) $P(A,B) = P(A) * P(B|A) = P(B) * P(A|B) \rightarrow \text{isolate this} \rightarrow P(A|B) = P(A) \frac{P(B|A)}{P(B)}$
 - ii) "Learning" Rule
 - iii) Think of film that is a clear representation of your taste (Hunger Games), then take The Avengers, belief that there is a 7% chance she has watched to avengers ($P(A)$), her choice (new information → B) was the Hunger Games, now there is a belief that there is a 15% chance that she has watched The Avengers **given new information**

$P(A) \rightarrow$ prior probability

$P(A|B) \rightarrow$ posterior probability