

Engaging Students in the Time of Business Analytics

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Perspective

Currently...

Teaching required undergrad regression

“Analytical” regression

Focus on answering business questions

Emphasize motivation/interpretation/action rather than “theory”

Software such as JMP, Excel, StatCrunch

3 sections, each with 120 students

1 TA

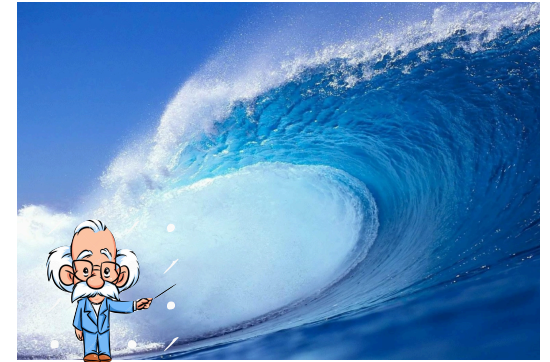
Gateway or barrier to later courses...

Business Analytics

Everywhere

Particularly in business schools

Elsewhere “Data Science”



At Wharton

New joint Business Analytics concentrations

Undergraduate

MBA

Attractive STEM field

Facilitates VISA

Soaring enrollment for many courses

Modern data analytics

R programming

Probability... one section @ 30 has grown to 3 @ 65

Challenges to Engagement

Scale

Attention span

Phones

Behavioral changes



Freshmen, sophomores

Don't know business: e.g. CAPM, elasticity, ...

Business Analytics has expanded variation

Growing subset of AP students adds yet more

Placement: Statisticians' view of Business Analytics is not the same as recruiters' view

Bus Analytics ≠ Applied Stat

Talk to those hiring Business Analytics grads

Data manipulation

Programming, scripting

Problem solving, willing to dive in

And, yes, statistical sense for data analysis

```
-----  
# Convolutional network example  
-----  
# placeholders for later data: x is naveled image, y is labels  
x = tf.placeholder("float", shape=[None, 784])  
y = tf.placeholder("float", shape=[None, 10])  
  
# utility functions for initializer and convolution  
def weight_variable(shape):  
    initial = tf.random_normal(shape, stddev=0.1)  
    return tf.Variable(initial)  
  
def bias_variable(shape):  
    initial = tf.constant(0.1, shape=shape)  
    return tf.Variable(initial)  
  
def conv2d(x, W):  
    return tf.nn.conv2d(x, W, [1, 1, 1, 1], padding='SAME')  
  
def max_pool_2x2(x):  
    return tf.nn.max_pool(x, [1, 2, 2, 1], padding='SAME', strides=[1, 2, 2, 1])
```

Wharton concentration

Joint concentration with “Decision Science”

OID

Deliberate distinction from Stat concentration

Probability course is not in the concentration!

Mixture: Data collection, Data analysis, Optimization

Lesson: Collaboration enhances marketable skills

Lack Business Knowledge

Undergraduate business program

Revised undergrad curriculum

First year exploration

Second and third year skills

Senior year capstone



Capstone course

Limited size, group project

Blend knowledge from various courses

Application field project

Hard to find, harder to sustain over time...

Lesson: More collaboration!

Engagement at Scale

Compelling examples are key

Distract from those phones as well...

Compelling for MBA \neq Compelling for Undergrad

Topical examples

Themselves

Contemporary issues

Software

Not teaching programming

At least not at the introductory level!

Jazzy animation

Examples that click

Student generated data

Number text messages, male vs female

Hard to manage (much less grade) at scale

projects

Substantive examples

Contemporary: Election surveys, sports

Climate change

Crime rates

Wage discrimination

“Eye candy” examples

Macroeconomic data

Election Surveys

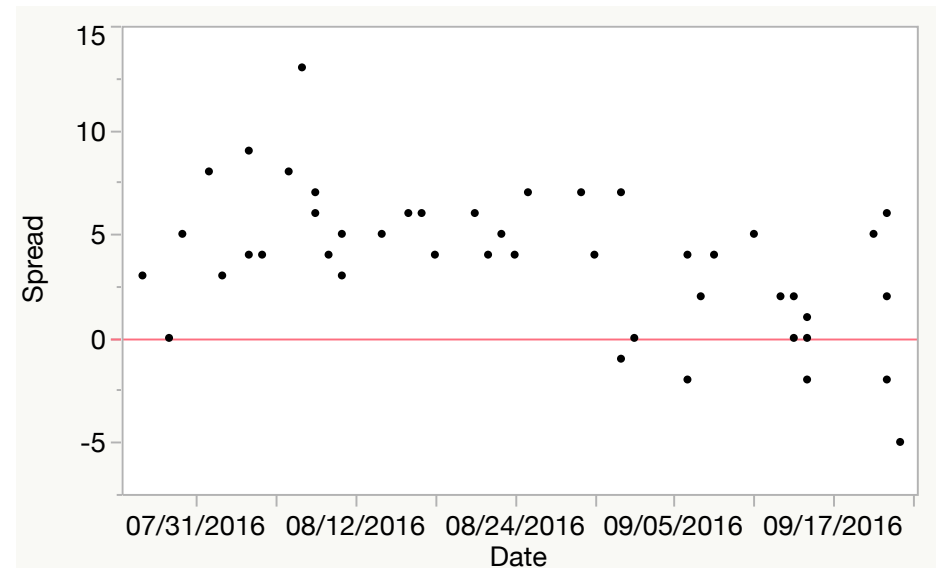
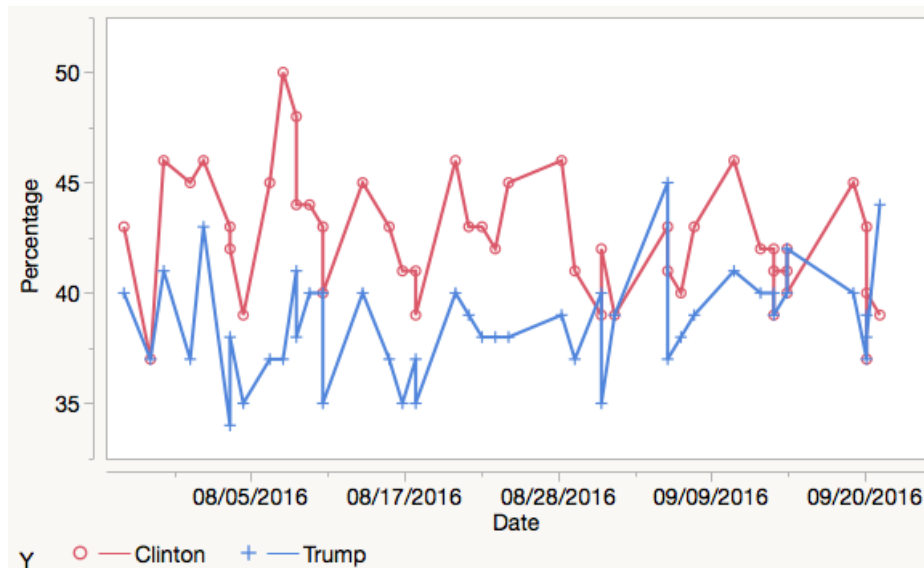
Topic

Data collection, survey design/execution

Sampling variation

Example

Surveys prior to recent election



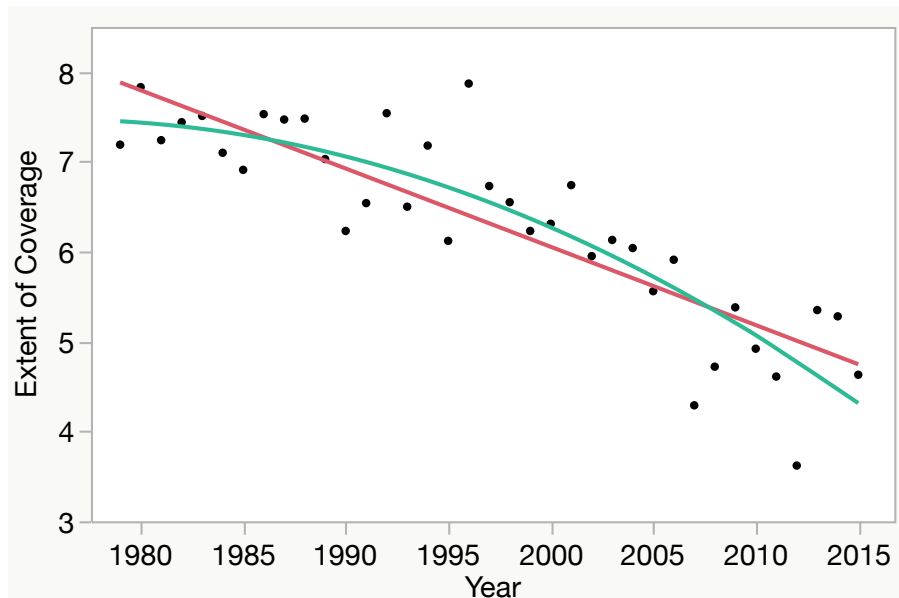
Climate Change

Topic

Polynomial trends, time series

Example

Arctic ice extent (million km²)



Linear or
Quadratic?

Assumptions?

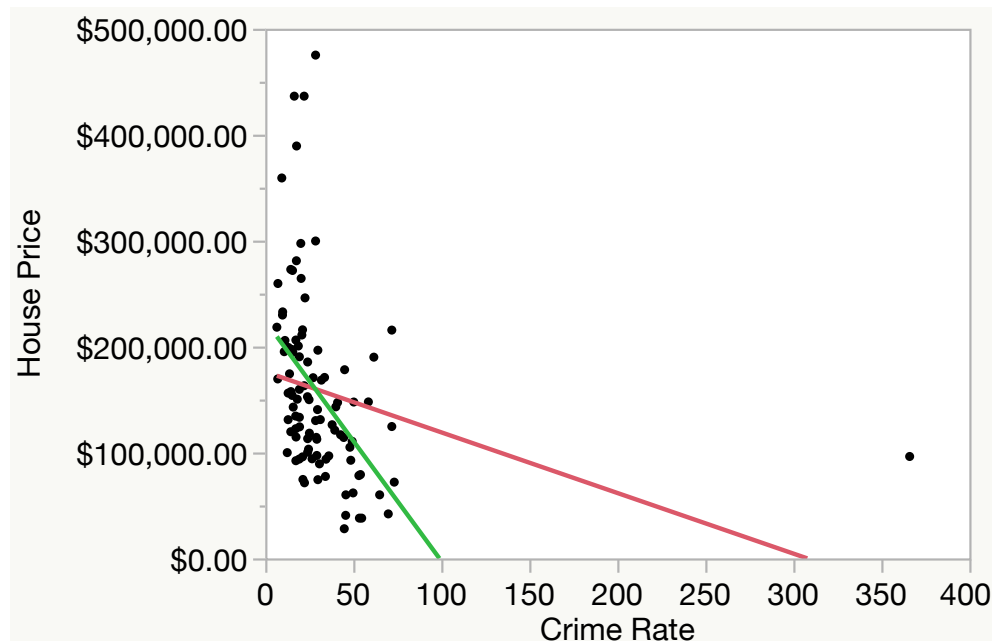
Crime Rates

Topic

Leveraged outliers

Example

Association between crime rates and housing values in communities in Philadelphia area



Where is that?
Why is it so different?

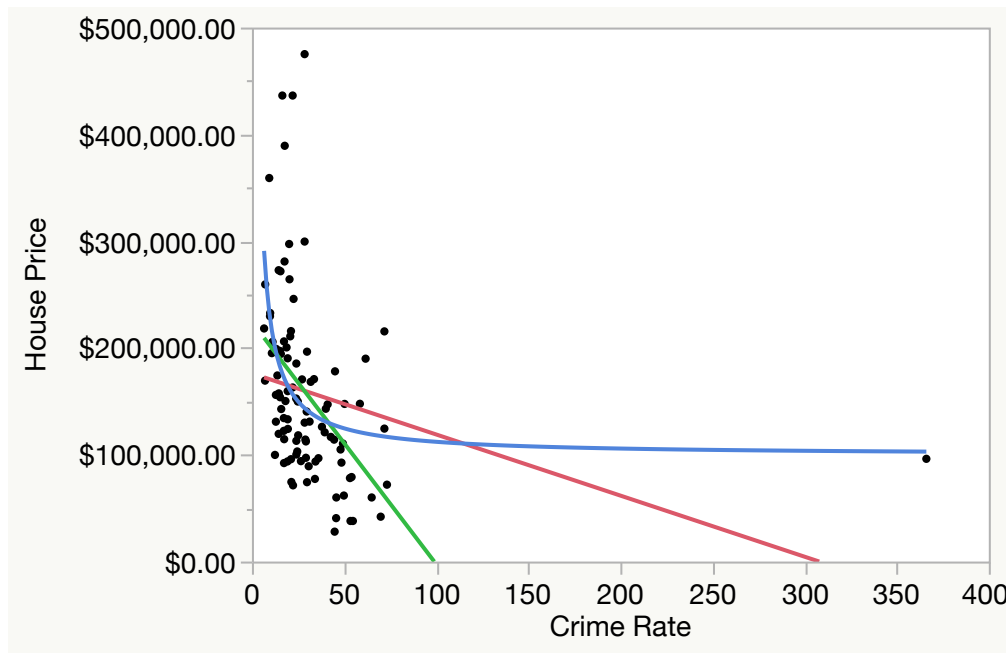
Crime Rates

Topic

Leveraged outliers, nonlinear trends

Example

Association between crime rates and housing values in communities in Philadelphia area



Where is that?
Why is it so different?

Linear?

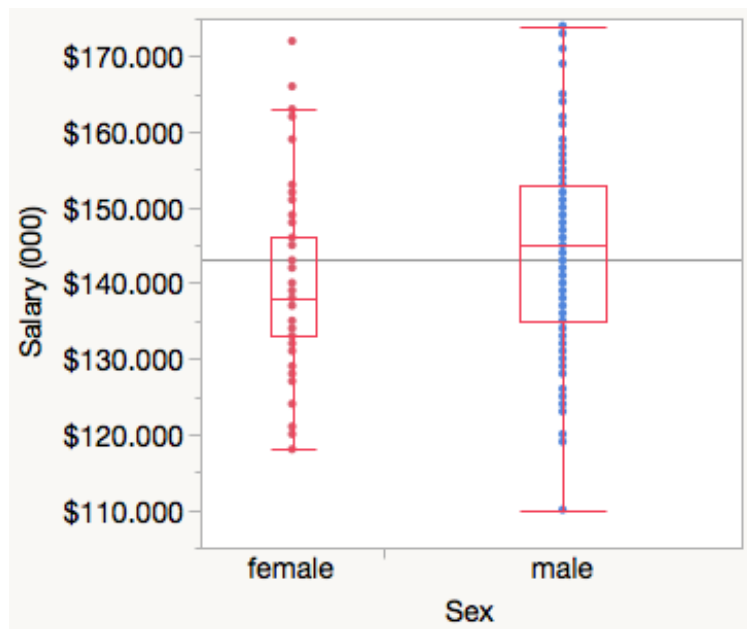
Wage Discrimination

Topic

Dummy variables in regression

Example

Adjusting for confounding effects in two-sample comparisons



male-female			
Assuming unequal variances			
Difference	4.67045	t Ratio	2.390108
Std Err Dif	1.95407	DF	122.4104
Upper CL Dif	8.53861	Prob > t	0.0184*
Lower CL Dif	0.80229	Prob > t	0.0092*
Confidence	0.95	Prob < t	0.9908

Wage Discrimination

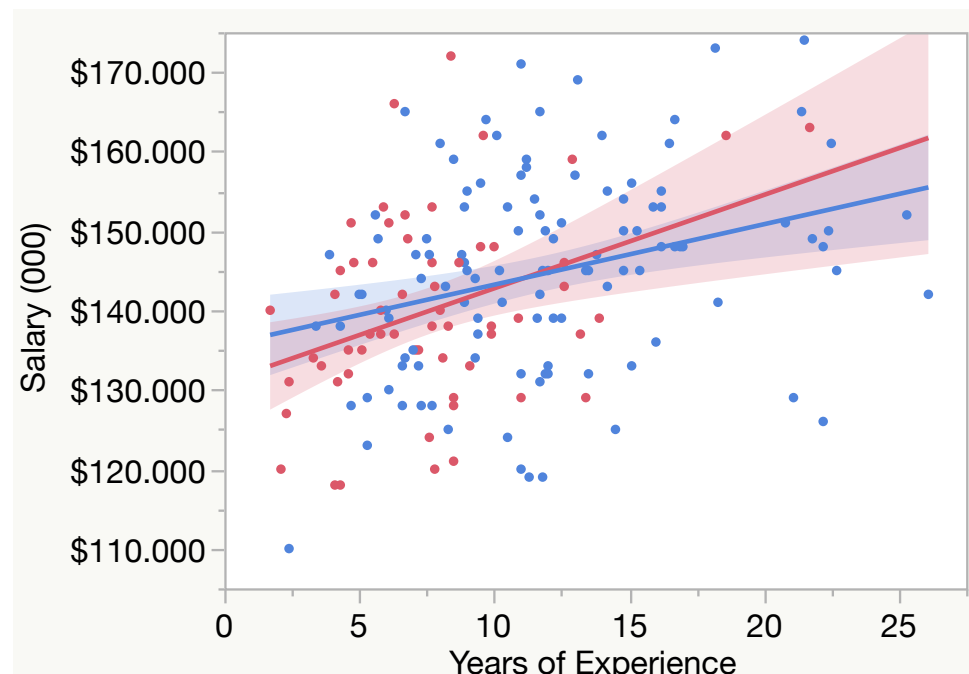
Topic

Dummy variables in regression

Example

Adjusting for confounding effects in two-sample comparisons

Knowing how to
use dummy
variable is
irrelevant if don't
know when!



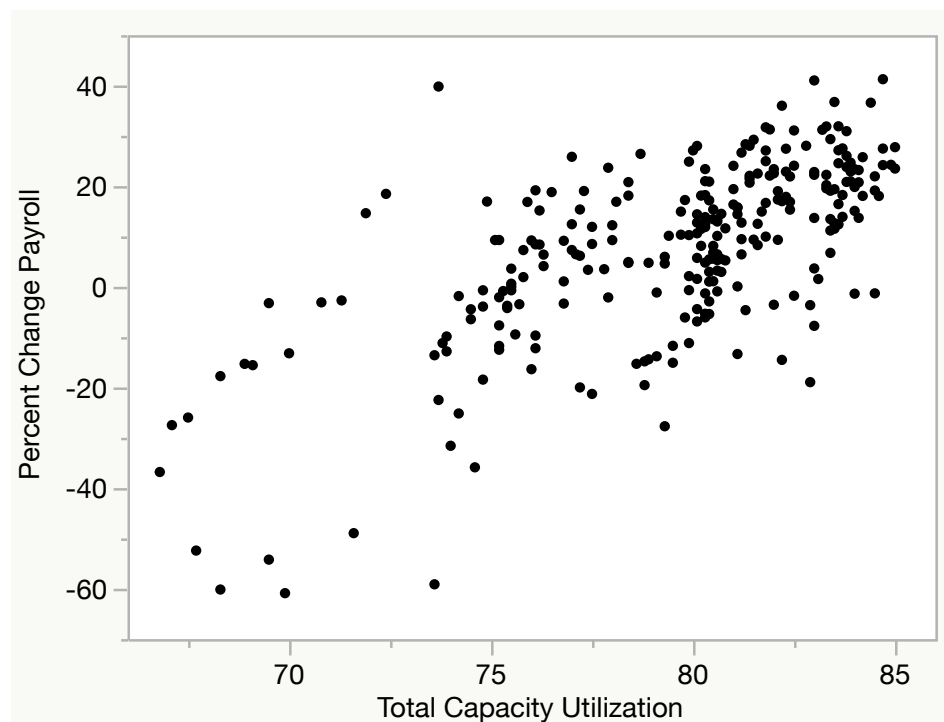
Macroeconomic Data

Macro data common in other courses

Why are there circular patterns in this plot?

Monthly macroeconomic data

Payrolls and capacity utilization



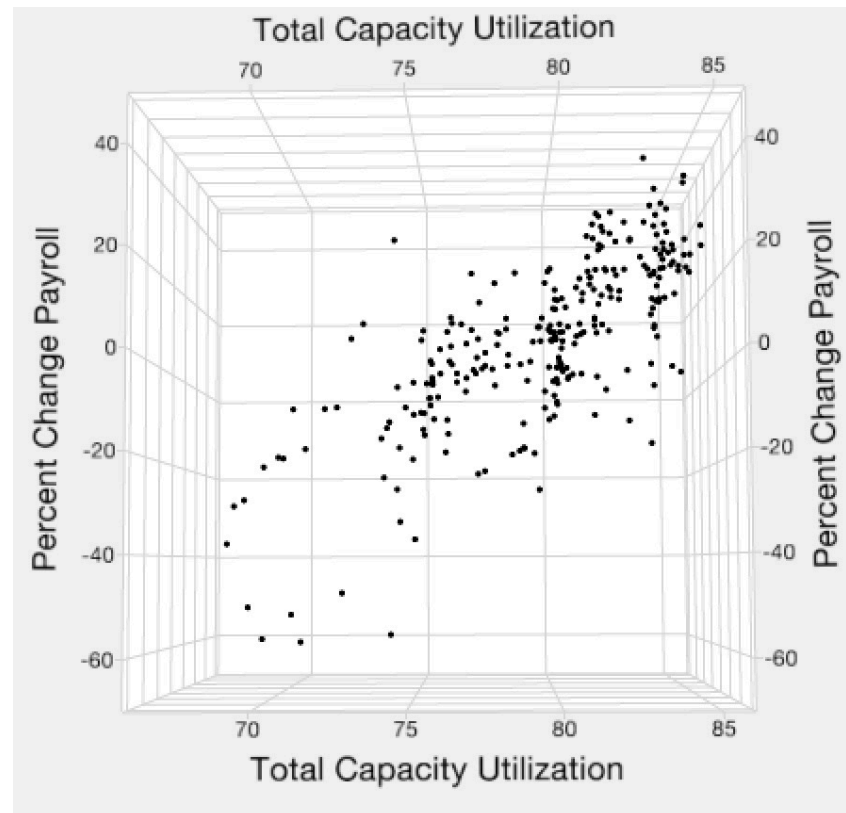
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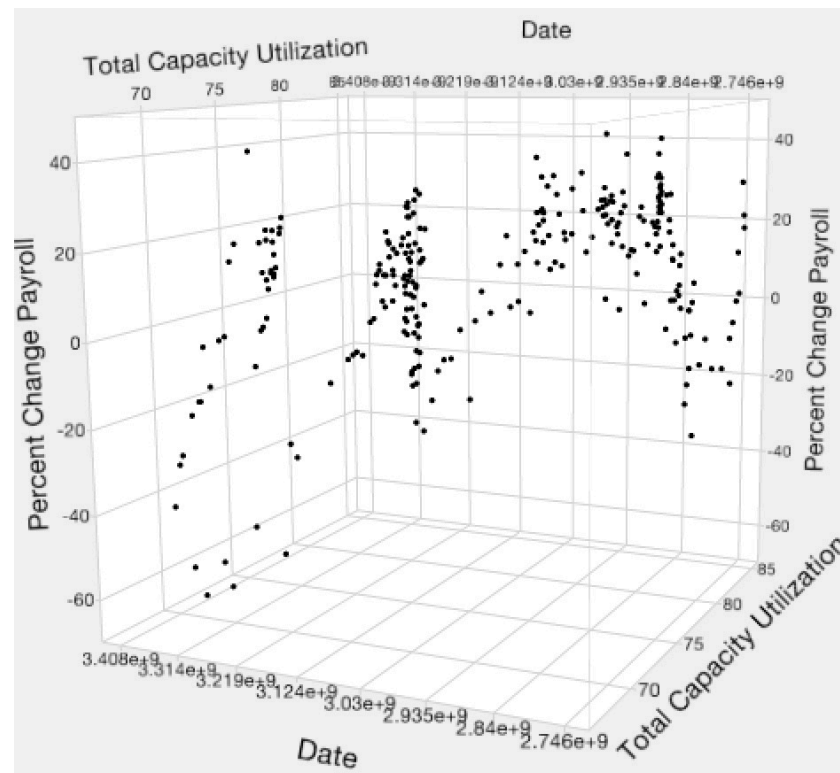
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Payrolls and capacity utilization

At some point to show that Stat actually leverages technology!



Discussion

Examples draw attention, curiosity

- Need to be topical to students

- Have to be fresh

Collaboration with other fields

- More important in age of “data science”

- Reach outside B School: computer science

Other thoughts

- Passion

- Respect