Under Used Statistical Pedagogical Ideas...

Some junk I've stumbled on that may help teaching stats and data analysis...

James (JD) Long

Here's my pitch...

Goal is to Kick Ass	Best Data	Teach Toy Models	Teach Meta Skills
Student retention is highest when a concept helps them do something they feel is useful. They don't want a linear progression of ideas that build incrementally. They want super powers to do something useful.	Fully controlled simulation OR Data students care about Nothing in between	Students who learn a theorem know one theorem. Students who can simulate a problem can back into MANY theorem's and gain better understanding and intuition.	"Learn to learn" Learn to embrace not knowing because you know the solution to ignorance. Ignorance is a solved problem.

So Who Am I?













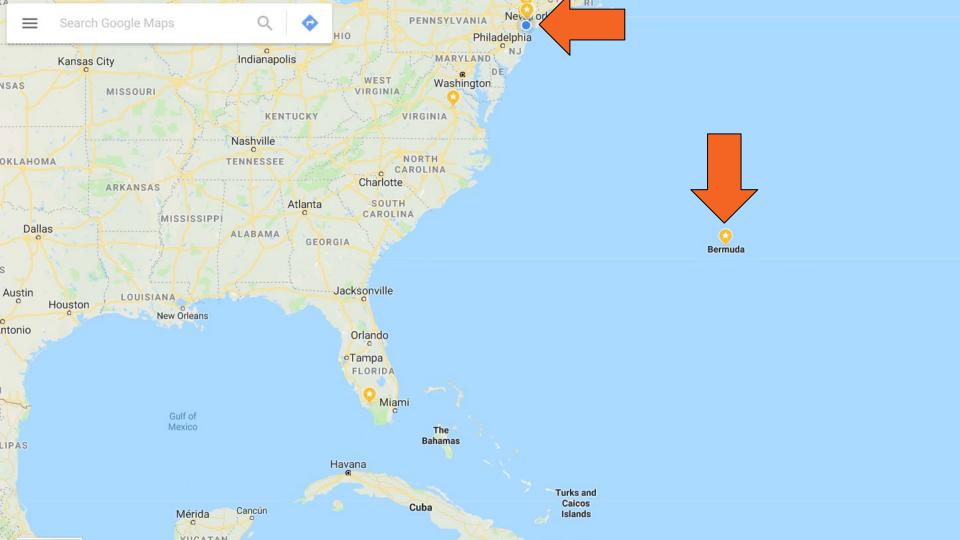














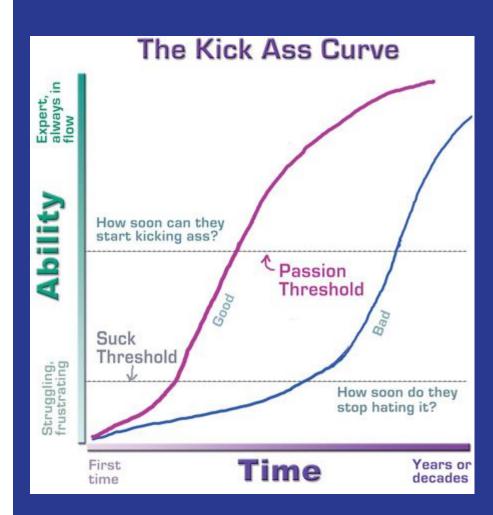
Nobody wants to learn analytics.
They want to learn to kick ass.

"The more knowledge and skill someone has, the more passionate they become, and the more passionate they become, the more they try to improve their knowledge and skills."

- Kathy Sierra

The Kick Ass Curve

Also borrowed from Kathy Sierra



How to Raid Fort Kickass

Build Motivation

Toddlers...

Toddlers learn to walk not because it's the next syllabus item. They learn because they want to carry two toys at once.

Get to the Good Stuff

Where are we going?

Illustrate what a concept will allow the student to **DO** in the future... DOing is more motivating than knowing

Make the Hard Bits Easier

Put the bowling bumpers up sometimes

It's OK to skip some ugly bits... then come back to them later if it builds motivation.

e.g. use 'Tidyverse' code in R to make R easier to use

Let's talk about example data...

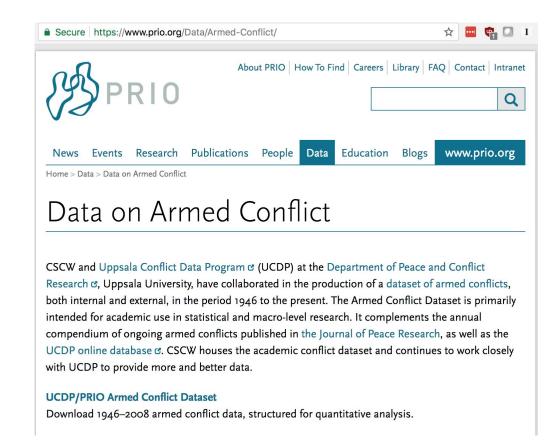
Fully Simulated Data

Data Students
Care About

What might learners care about?

Actual business data that lead to a story...

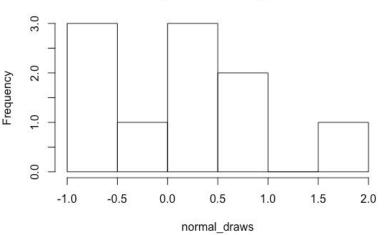
Conflict data?
The Peace Research
Institute Oslo (PRIO)



Fully Simulated Data...

```
sample size <- 10
# simple illustration of random draws
normal draws <- rnorm(sample size, mean=0,
sd=1)
mean (normal draws)
                                Histogram of normal_draws
hist (normal draws)
```

[1] 0.1322028



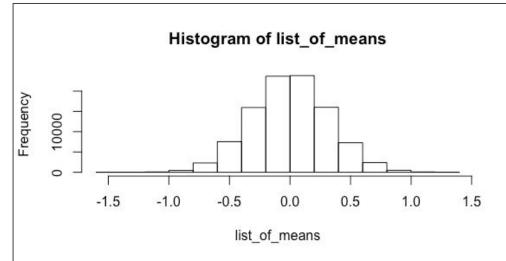
what's the distribution of the mean measurements as we
do this over and over?
list of means <- array()</pre>

```
for (i in 1:times_to_loop) {
  normal_draws <- rnorm(sample_size, mean=0, sd=1)
  list_of_means[i] <- mean(normal_draws)</pre>
```

7 Lines of code... And well on our way to backing into Student T test, sqrt(n) intuition

hist(list of means)

times to loop <- 100000



Building Toy Models: Expansion of Simulated Data

```
## let's play with a regression now
draws <- 1000
set.seed(2)
# create a DF with 3 columns, 1000 rows of random standard normal draws
random regression <- data.frame(replicate(3,rnorm(draws)))</pre>
#calculate the dependent variable Y
random regression %>% mutate(
    e = rnorm(draws, 0, 1), ## better add some error noise
    Y = 2 * X1 + 3 * X2 + 4 * X3 + e
) -> random regression
# build a linear regression
model <- lm( Y ~ X1 + X2 + X3, data=random regression )</pre>
summary(model)
```

```
Call:
lm(formula = Y \sim X1 + X2 + X3, data = random regression)
Residuals:
   Min 10 Median 30 Max
-3.2144 -0.6782 0.0100 0.6499 3.1942
Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.02382
                0.03135 0.76 0.448
         1.93678 0.03079 62.90 <2e-16 ***
X1
X2.
         3.04803 0.03144 96.95 <2e-16 ***
         Х3
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

Residual standard error: 0.9872 on 996 degrees of freedom Multiple R-squared: 0.9689, Adjusted R-squared: 0.9688 F-statistic: 1.036e+04 on 3 and 996 DF, p-value: < 2.2e-16

Other Toy Models?

Actual Experience:

If we have 30 observations from a lognormal distribution, what's our confidence around the 90% percentile tail measurement? What about the 50% percentile?

Meta Skills:

Ultimately we need only teach one skill:

How to learn something we don't already know.

Top Technical Meta Skills

- How to create a reproducible example
- How to ask a question
- How to explain a problem
- How to query Google
- How to RTFM
- How to document a process
- How to pick the right tool
- How good is good enough
- Learn that "design patterns" exist

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https://github.com/CerebralMastication/WestPointPresentation