# Using data to confuse and deceive

#### Matthew Brett

#### Handout

Available at https://github.com/matthew-brett/lans-workshop/blob/master/data\_confuse\_deceive\_handout.pdf

#### How to lie with statistics

If you can't prove what you want to prove, demonstrate something else and pretend that they are the same thing. In the daze that follows the collision of statistics with the human mind, hardly anybody will notice the difference.

(Huff 1954)

# "Research Excellence" Framework

Over thirty years the RAE / REF has supported a sustained improvement in the quality and productivity of the UK research base. It is used by universities to attract students, staff and external funding.

Research Excellence Framework (REF) review: Building on success and learning from experience by Lord Nicholas Stern.

# What is improvement?

	1947–66	1967-86	1987-2006
US	50	88	126
UK	20	25	9

Number of science Nobel prizes by country and time period. REA / REF started in 1986.

(Charlton 2007)

## What is improvement?

In contrast to the picture of long term decline in Nobel-prize-winning revolutionary science; UK and European scientific production (also that of Chinese science) is probably catching up with the USA in terms of scientometric measures such as numbers of publications and citations.

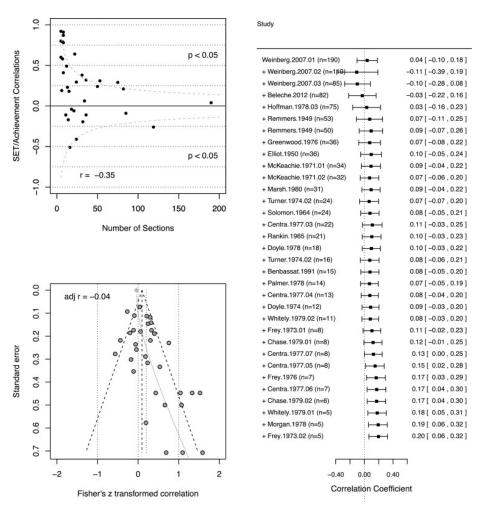
(Charlton 2007)

# "Teaching Excellence" Framework

- student satisfaction using the teaching on course, assessment and feedback and academic support scales from the National Student Survey;
- retention using [Higher Education Statistics Agency] UK Performance Indicators;
- proportion in employment in further study using 6 month [Destination of Leavers from Higher Education Survey].

UK government Higher education: success as a knowledge economy - white paper

# Student Evaluations do not measure teaching effectiveness



(Uttl, White, and Gonzalez 2017)

# Summary of meta analysis

The reported correlations between [Student Evaluations of Teaching] ratings and learning are completely consistent with randomly generating correlations from the population correlation with rho = 0 and applying publication selection bias.

(Uttl, White, and Gonzalez 2017)

# What do student ratings measure?

- the grade the student expects to get;
- the subject being taught;
- whether the instructor is white and male
- biscuits

# Expected grade

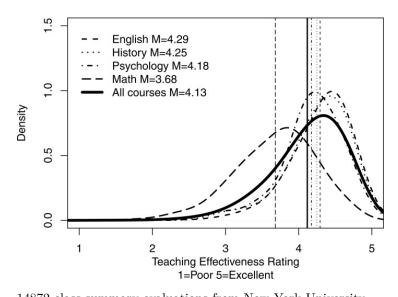
Table 7. Average correlation between SET and interim grades

	$ar{ ho}$	р
Overall	0.16	0.00
History	0.32	0.00
Political institutions	-0.02	0.61
Macroeconomics	0.15	0.01
Microeconomics	0.13	0.03
Political science	0.17	0.02
Sociology	0.24	0.00

p-values are one-sided.

(Boring, Ottoboni, and Stark 2016)

# Subject being taught



 $14872~{\rm class}$  summary evaluations from New York University (Uttl and Smibert 2017)

#### Whether the instructor is male

**Table 8.** Mean ratings and reported instructor gender (male minus female).

	Difference in means	Nonparametric <i>p</i> -value	MacNell et al. <i>p</i> -value
Overall	0.47	0.12	0.128
Professional	0.61	0.07	0.124
Respectful	0.61	0.06	0.124
Caring	0.52	0.10	0.071
Enthusiastic	0.57	0.06	0.112
Communicate	0.57	0.07	NA
Helpful	0.46	0.17	0.049
Feedback	0.47	0.16	0.054
Prompt	0.80	0.01	0.191
Consistent	0.46	0.21	0.045
Fair	0.76	0.01	0.188
Responsive	0.22	0.48	0.013
Praise	0.67	0.01	0.153
Knowledge	0.35	0.29	0.038
Clear	0.41	0.29	NA

p-values are two-sided.

(Boring, Ottoboni, and Stark 2016)

#### **Biscuits**

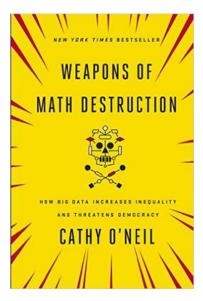
The cookie group evaluated teachers significantly better than the control group (113.4 pm 4.9 versus 109.2 pm 7.3; p = 0.001, effect size 0.68). Course material was considered better (10.1 pm 2.3 versus 8.4 pm 2.8; p = 0.001, effect size 0.66) and summation scores evaluating the course overall were significantly higher (224.5 pm 12.5 versus 217.2 pm 16.1; p = 0.008, effect size 0.51) in the cookie group.

(Hessler et al. 2018)

#### For discussion

- Given the data here, why did the government want to use student ratings to evaluate teaching?
- Imagine I propose an alternative rating of teaching excellence, which is a random number between 1 and 10. Make arguments for preferring a metric based on student evaluations.
- If you could decide how to evaluate teaching, what do you propose?

#### Algorithms and public policy



(O'Neil 2016)

# Risk scores for re-offending

# Machine Bias

There's software used across the country to predict future criminals. And it's biased against blacks.

by Julia Angwin, Jeff Larson, Surya Mattu and Lauren Kirchner, ProPublica May 23, 2016

Full analysis at https://github.com/propublica/compas-analysis

# COMPAS

"Correctional Offender Management Profiling for Alternative Sanctions" Proprietary algorithm Marketed by Northpointe

# Risk score officer questions

8.	How many prior misdemeanor assault offense arrests (not sex, or domestic violence) as an adult? O 0 O 1 O 2 O 3+
9.	How many prior family violence offense arrests as an adult? O 0 O 1 O 2 O 3+
10.	How many prior sex offense arrests (with force) as an adult? O 0 O 1 O 2 O 3+
11,	How many prior weapons offense arrests as an adult? $\bigcirc$ 0 $\bigcirc$ 1 $\bigcirc$ 2 $\bigcirc$ 3+
12.	How many prior drug trafficking/sales offense arrests? $\bigcirc$ 0 $\bigcirc$ 1 $\bigcirc$ 2 $\bigcirc$ 3+

COMPAS questionnaire

# Risk score offender questions

24.	In the last 12 months before this incarceration, how often did you have contact with your family (may be in person, phone, mail)?
	O No family O Never O Less than once/month O Once per week O Daily
25.	In the last 12 months before this incarceration, how often did you move? $\bigcirc$ Never $\bigcirc$ 1 $\bigcirc$ 2 $\bigcirc$ 3 $\bigcirc$ 4 $\bigcirc$ 5+
26.	Did you have a regular living situation prior to your current incarceration (an address where you routinely stayed and could be reached)?
	O No O Yes
27.	How long had you been living at your last address prior to this incarceration? O 0-5 mo. O 6-11 mo. O 1-3 yrs. O 4-5 yrs. O 6+ yrs.
28.	Was there a telephone at this residence (a cell phone is an appropriate alternative)? $\bigcirc$ No $\bigcirc$ Yes

COMPAS questionnaire

#### Risk score results

# Name: Class3, Jessie SSN: Offender #: 01cr57 Date of Birth: 06/19/1977 Date of Screening: 08/14/2006 Comment: Overall Risk Potential Decile Score Violence 1 Recidivism 3 Failure to Appear 4 Community Non-compliance 4 Criminal History Factors Criminal Involvement 3 History of Non-Compliance 1 Name: Class3, Jessie SSN: Offender #: 01cr57 Date of Screening: 08/14/2006 Overall Risk Potential Decile Score Medium Risk High Risk H

COMPAS results

# Background - a crash course on machine learning

See https://www.inferentialthinking.com/chapters/17/Classification

# Prediction errors by race

Black Defendants		White Defendants			
	Low	High		Low	High
Survived	990	805	Survived	1139	349
Recidivated	532	1369	Recidivated	461	505
FP rate: 44.85			FP rate: 23.45		
FN rate: 27.99			FN rate: 47.72		
PPV: 0.63			PPV: 0.59		
NPV: 0.65			NPV: 0.71		
LR+: 1.61			LR+: 2.23		
LR-: 0.51			LR-: 0.62		

See this description of the analysis.

#### Garbage in, Gospel out

From Garbage in, garbage out

#### Assessing risk scores

Although these measures were crafted with the best of intentions, I am concerned that they inadvertently undermine our efforts to ensure individualized and equal justice.

Eric Holder, US Attorney General, 2014 (quoted in Propublica).

#### For discussion

- Given the importance of the risk scores, and their potential for bias, why was there so little study of their performance?
- What would be your recommendation for the use of the Northpointe system studied here?
- What recommendations would you make, to a state that was considering using a system like this? What resources would you provide to help them?

#### Is this the end?

Yes, it's the end of the workshop.

All material for these slides at https://github.com/matthew-brett/lans-workshop

#### References

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