## LIFE AND DEATH DECISIONS:

# TESTING DATA SCIENCE

## WHY TESTING IS IMPORTANT







Owner video of Autopilot steering to avoid collision with a truckm.youtube.com/watch?feature=...

5:34 PM - Apr 17, 2016

 $\bigcirc$  5,604  $\bigcirc$  2,517 people are talking about this





Blog Videos Press Customer Storie

#### A Tragic Loss

The Tesla Team . June 30, 2016

Be the firs news, eve

example

We learned yesterday evening that NHTSA is opening a preliminary evaluation into the performance of Autopilot during a recent fatal crash that occurred in a Model S. This is the first known fatality in just over 130 million miles where Autopilot was activated. Among all vehicles in the US, there is a fatality every 94 million miles. Worldwide, there is a fatality approximately every 60 million miles. It is important to emphasize that the NHTSA action is simply a preliminary evaluation to determine whether the system worked according to expectations.

Following our standard practice, Tesla informed NHTSA about the incident immediately after it occurred. What we know is that the vehicle was on a divided highway with

https://www.tesla.com/blog/tragic-loss

## SINCE THEN

- 23 March 2018 Tesla X crashed into a "poorly maintained crash attenuator" <sup>1</sup>
- 18 March 2018 Uber Ford Fusion hit a pedestrian<sup>2</sup>
- 1. https://www.theguardian.com/technology/2018/marcar-crash-autopilot-mountain-view
- 2. https://www.theguardian.com/technology/2018/maiself-driving-car-kills-woman-arizona-tempe

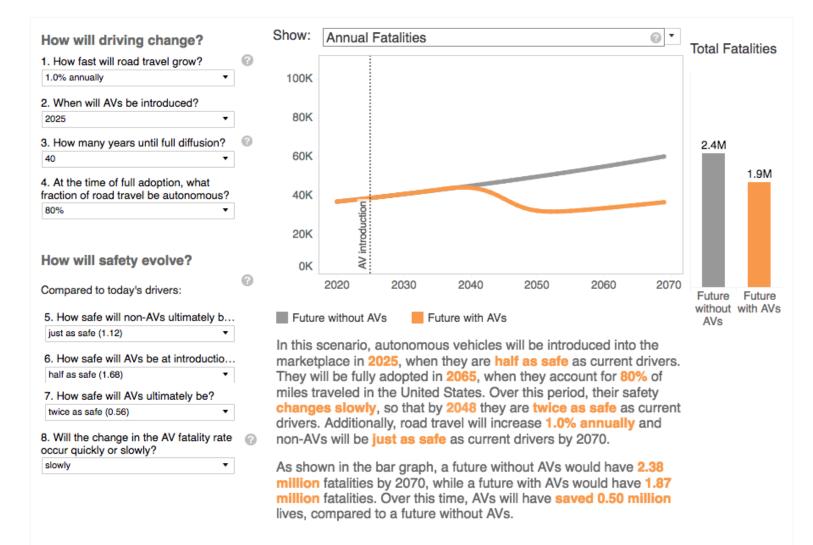
#### CONTEXT

 Approximately 37,000 deaths per year source on US roads, where 93% of them can be attributed to human error source

ACTIVITY

SIGN IN

RAND - Profile



## RAND Corporation via Tableau

## **SENSATIONALISM?**

## CORRECTIONAL OFFENDER MANAGEMENT PROFILING FOR ALTERNATIVE SANCTIONS (COMPAS)

#### 2016

- Borden arrested for stealing an \$80 childrens bike.
- Rated as high risk

--

- Prater arrested for stealing \$86 worth of tools from Home Depot.
- Rated as low risk.
- Julia Angwin, Jeff Larson, Surya Mattu and Lauren Kirchner, ProPublica May 23, 2016
- Inspired by Cody Marie Wild via HackerNoon

## Two years later:

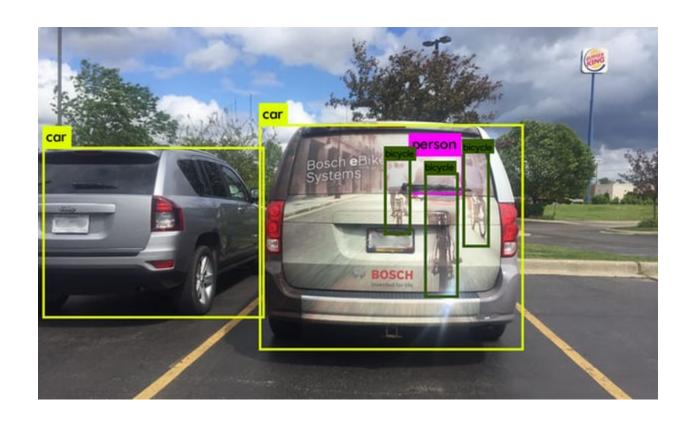
• Borden was not charged with any other crimes.

--

 Prater is serving an eight-year prison term for stealing thousands of dollars' of electronics from a warehouse. • Borden was an African American Female

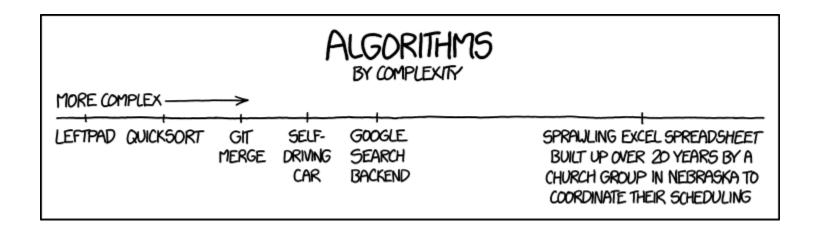
• Prater was a Caucasian Male

## **TESTING DURING TRAINING**



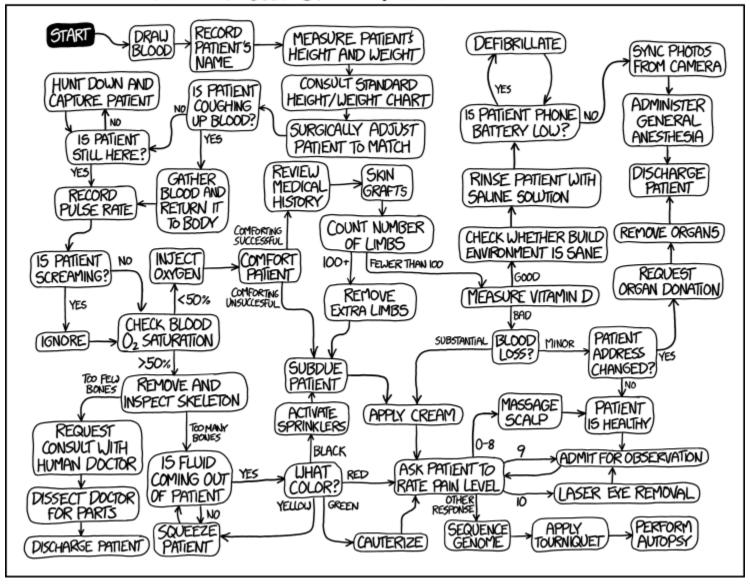
The Guardian, Assume self-driving cars are a hacker's dream? Think again, 30 Aug 2017

## COMPLEXITY

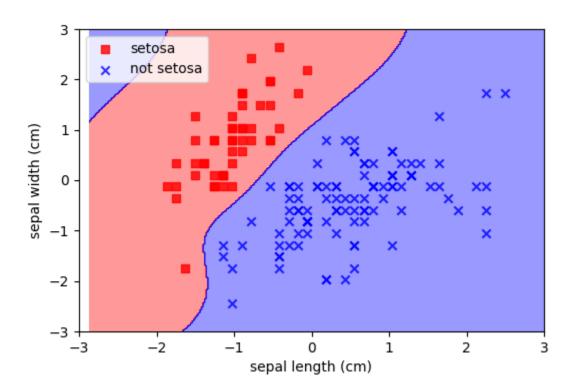


XKCD - Algorithms

#### A GUIDE TO THE MEDICAL DIAGNOSTIC AND TREATMENT ALGORITHM USED BY IBM'S WATSON COMPUTER SYSTEM



- XKCD Watson Medical Algorithm
- Explanation of XKCD Watson Medical Algorithm



## MORE DATA IS BETTER

## **METRICS**

- Business metrics
- Technical metrics

## **TYPES OF PROBLEM**

- Regression
- Classification
- Clustering

## **CONFUSION MATRIX**

A confusion matrix is a table showing the degree of class confusion.

	actual positive	actual negative
predict true	true positive	false positive
predict false	false negative	true negative

## UNBALANCED CLASSES

Very common to have classes that are very rare.

Skews of 1: 100 are common in fraud detection, and skews of greater than  $1:10^6$  have been reported in other applications<sup>1</sup>.

I) Attenberg, J., & Provost, F. (2010, July). Why label when you can search?: alternatives to active learning for applying human resources to build classification models under extreme class imbalance.

	positive	negative
yes	0	0
no		49

$$\frac{accuracy}{accuracy} = \frac{\text{Number of correct decisions made}}{\text{Total number of decisions made}} =$$

## After balancing classes:

	positive	negative
yes	0	0
no	25	25
ac	$curacy = \frac{25}{50}$	

## NUMERICAL MODEL EVALUATION

 Every decision you make has some impact on the business.

We need to talk in the same language as the business. Profit/Loss.

#### **EXPECTED VALUE**

## Marketing example:

- true positive: £50 £9 = £41.
- true negative: Did not send, would not buy.
- false positive: Sent, did not buy: -£9
- false negative: Did not send, but would have bought. £0, but opportunity cost?

We can represent this cost benefit data,  $b(\mathbf{x})$ , in a matrix:

	p	n
Y	41	-9
N	0	0

## VISUAL MODEL EVALUATION

• Numerical scores are necessary for optimisation. But visual representations are far more intuitive.

## RANKING

Most algorithms output some kind of score

For example:

- Probabilities
- Distances
- Errors

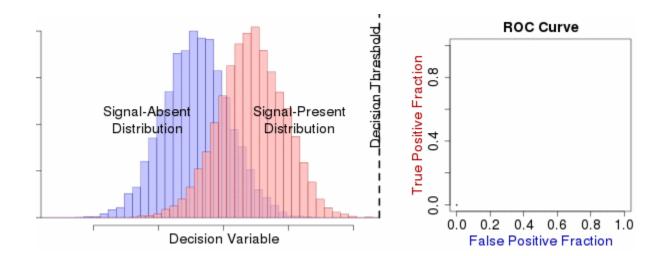
We can alter the *threshold* of that score to generate a new confusion matrix.

#	С	Score	1.0
1	Р	0,9	1,0
2	P	0,8	0,9
3	N	0,7	0,5
4	Р	0,6	0,8
5	P	0,55	
6	P	0,54	0,7
7	N	0,53	
8	N	0,52	0,6
9	P	0,51	
10	N	0,505	0,5
11	P	0,4	0.4
12	N	0,39	0,4
13	Р	0,38	0,3
14	N	0,37	
15	N	0,36	0,2
16	N	0,35	
17	Р	0,34	0,1
18	N	0,33	
19	Р	0,3	0,0 4
20	N	0,1	0,0 0,1 0,2 0,3 0,4 0,5 0,6 0,7 0,8 0,9 1,0

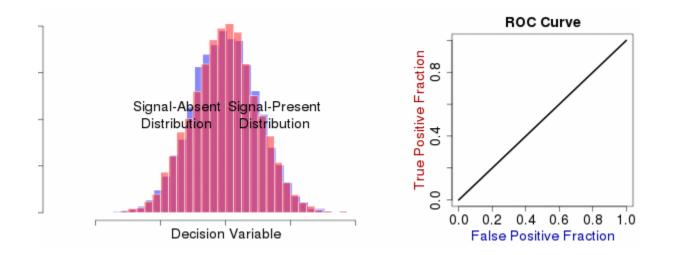
http://mlwiki.org/index.php/ROC\_Analysis

## **ROC PLOTS AND CURVES**

• Receiver Operating Curves (ROC) plot the false positive rate vs. the true positive rate



http://davidgbrown.co/

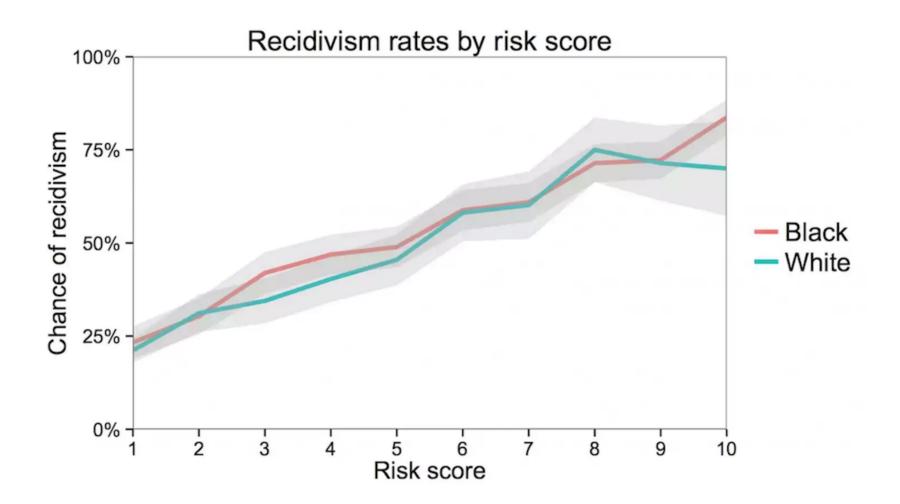


http://davidgbrown.co/

#### **WARNING!**

Many types of curves are not balanced!!

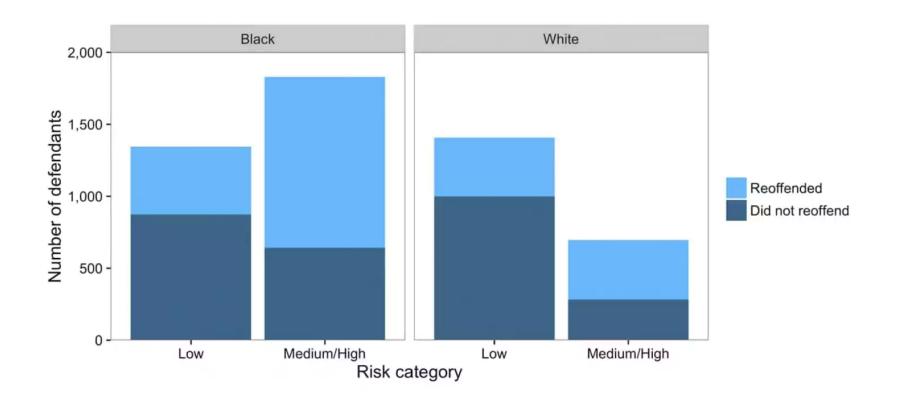
### IS COMPAS FLAWED?



Sam Corbett-Davies, Emma Pierson, Avi Feller and Sharad Goel, Washington Post

## ProPublica's definition of "fairness" focused on prisoners that did not reoffend.

	Caucasian	African American
Higher Risk, Didn't Re-Offend	23.5%	44.9%
Lower Risk, Did Re- Offend	47.7%	28.0%
ProPublica		



Sam Corbett-Davies, Emma Pierson, Avi Feller and Sharad Goel, Washington Post

#### Caucasian defendants

Accuracy: 0.67

• False positive rate: 0.23

ProPublica Data

#### African American defendants

Accuracy: 0.64

False positive rate: 0.45
 ProPublica Data

- Accuracy is similar for each risk category
- There are twice as many African American reoffenders

• • •

 There are twice as many "false positives" for African Americans

- Balance groups over scores (calibration)
- Balance groups over "not-reoffend" (negative class balance, false positive rate)
- Balance groups over "reoffended" (positive class balance, false negative rate)

#### Choose I

Inherent Trade-Offs in the Fair Determination of Risk Scores, Jon Kleinberg, Sendhil Mullainathan, Manish Raghavan

- Cost/Benefit Analysis and optimise for expected value?
  - Does parole work?
  - Should everyone be treated fairly?
  - Should offenders be treated fairly?
  - Is it more important to keep prospective criminals off the street?
- Underlying problems
  - Why is there so much skew in the base rate?

- More questions hidden in the data
  - Differences between Men and Women
  - Similar problems with people of Hispanic descent
  - So many confounders (violence, drugs, background, etc.)
- Sampling problems
  - Can't sample everyone. Are samples representative?
- Local effects
  - Is it reasonable to assume one model for the whole of the US?

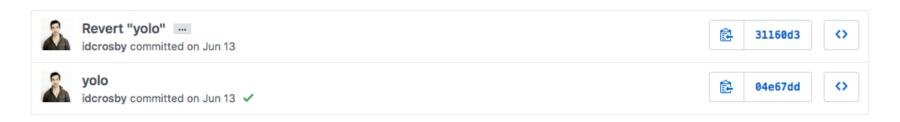
We found that sometimes people's names or dates of birth were incorrectly entered in some records - which led to incorrect matches between an individual's COMPAS score and his or her criminal records. We attempted to determine how many records were affected. In a random sample of 400 cases, we found an error rate of 3.75 percent (CI: +/- 1.8 percent).

# WHAT DOES THIS MEAN FOR ENGINEERS?

• Fallout



- Be careful
- You are biased
- Your data are biased
- Ask others to break it
- Visualise all the things
- Don't rely on single metrics
- Communicate performance (expected value framework)
- Be upfront with weaknesses

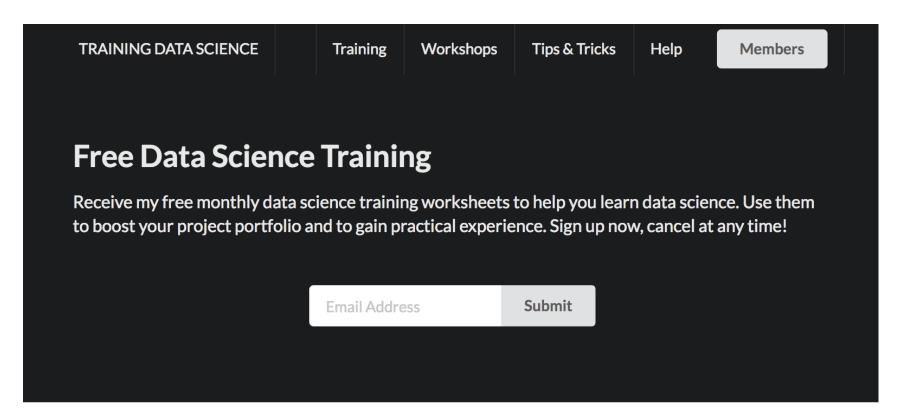


- Stream or batch
- Regression
- Clustering
- Testing in production/monitoring
- Feedback
- Online/offline differences
- Deployment
- Provenance
- Snapshotting
- Testing types (model/data/regression/etc.)
- Get the easy things right

## Data Science **Training, Consultancy, Development**

- OrPhilWinder
  - DrPhilWinder
- https://WinderResearch.com
- phil@WinderResearch.com

### https://TrainingDataScience.com



#### Why I Created Training Data Science

I'm tired of click-bait. I don't agree with the use of the word AI. I wanted to create a place where people could learn data science without the hype. This is for people that want to learn data

@DrPhilWinder | WinderResearch.com