

Pittsburgh, PA April 18-19, 2018

## Analyzing Systemic Racial Disparities With Statistical Learning Models and HMIS Data

Clayton Aldern

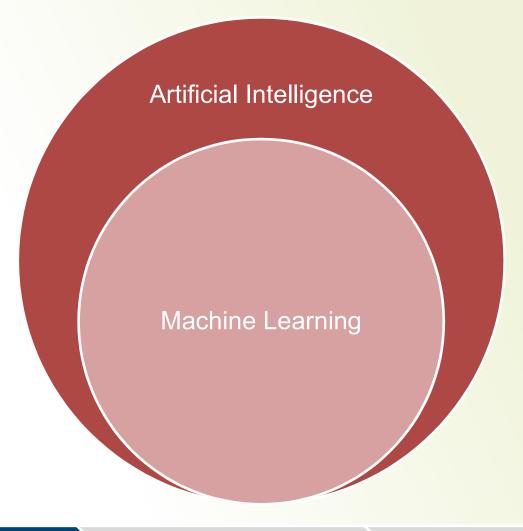
Pierce County, Washington | caldern@co.pierce.wa.us



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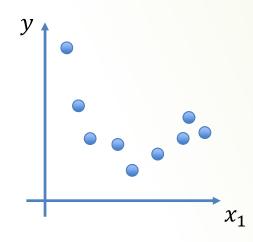
# Alternate title: Machine learning isn't special, but it can be useful

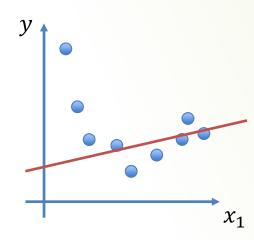
- I. What is a statistical learning model?
- II. Case study
- III. Tools

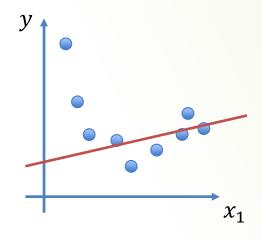


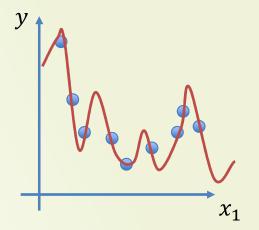
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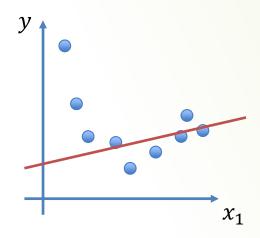
# We say a machine "learns" if it improves its performance P on a task T, given some experience E

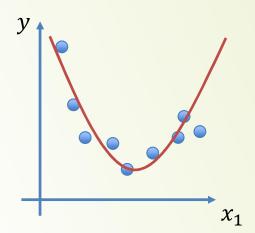


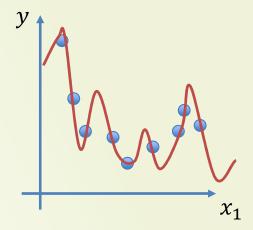




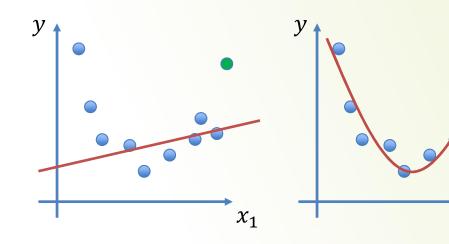


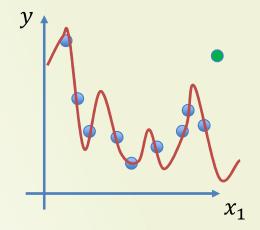






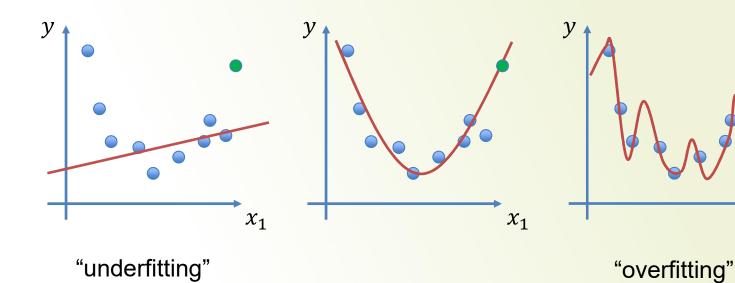
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 $x_1$ 

 $x_1$ 



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## What are the goals of model selection?

- Construct a good predictor. (Values of model coefficients are irrelevant.)
- II. Give causal interpretations of the factors (and determine which variables are "important").

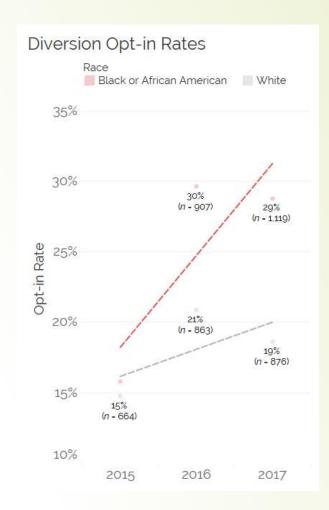
(Bickel and Li, 2006)

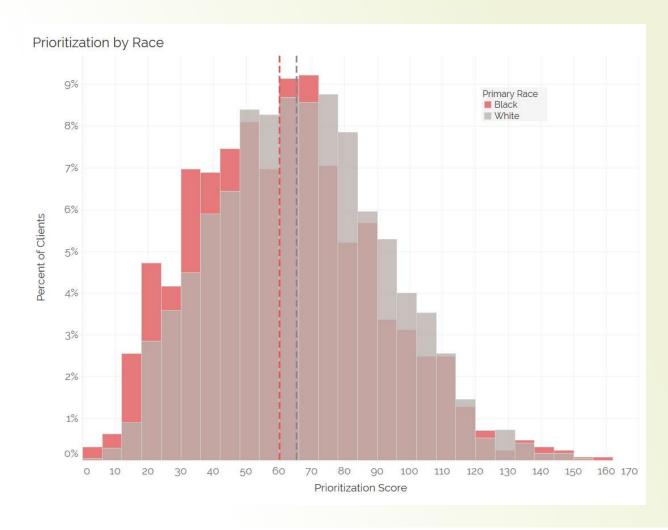
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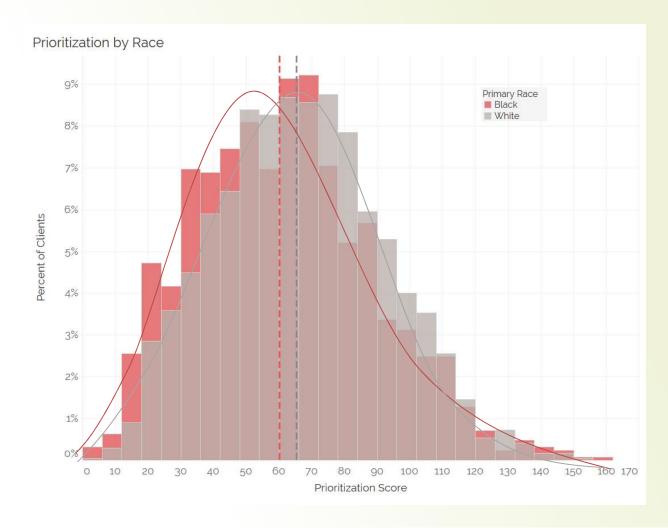
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### **Motivating observations**

I. Background II. Case Study III. Tools







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II. Case Study

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# Which assessment questions and responses are significantly correlated with race?

. Background III. Case Study III. Tools

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# Which as sea ment cuestions and responses are sonificantly correlate with race

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# Can we predict someone's race, given only their assessment responses?

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Race	Age	Gender	•••	Medical conditions in household
White	54	Male		4+
Black	32	Female		2
Black	33	Male		1
White	35	Female	•••	3

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Race	Age	Gender		medcon 1	medcon 2	medcon 3	medcon 4+
1	54	1		0	0	0	1
0	32	0	•••	0	1	0	0
0	33	1		1	0	0	0
		•••	•••	•••	•••	•••	•••
1	35	0		0	0	1	0

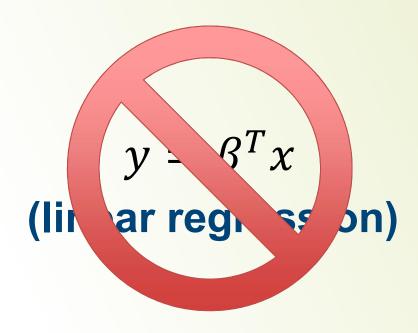
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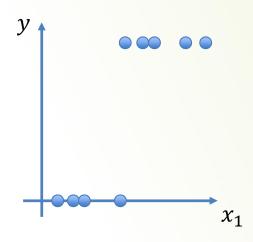
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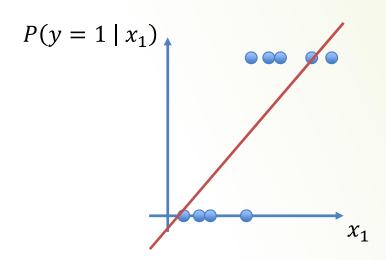
У	X <sub>1</sub>	X <sub>2</sub>	•••	X <sub>n-3</sub>	X <sub>n-2</sub>	X <sub>n-1</sub>	X <sub>n</sub>
1	54	1	•••	0	0	0	1
0	32	0	•••	0	1	0	0
0	33	1	•••	1	0	0	0
•••	•••	•••	•••	•••	•••	•••	•••
1	35	0		0	0	1	0

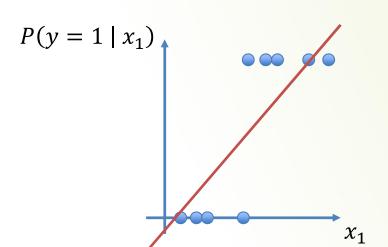
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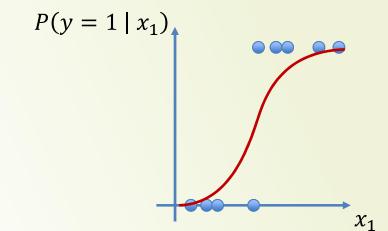
$$y = \beta^T x$$
 (linear regression)





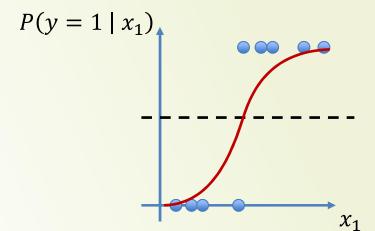






$$P(y = 1 \mid x_1)$$

$$x_1$$



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$$P(y = 1 \mid x) = h_{\beta}(x) = \frac{1}{1 + e^{(-\beta^T X)}}$$

$$\log\left(\frac{P(y=1\mid x)}{1-P(y=1\mid x)}\right) = \beta^T X$$

(logistic regression)

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$$P(y = 1 \mid x) = h_{\beta}(x) = \frac{1}{1 + e^{(-\beta^T X)}}$$

$$\log\left(\frac{P_{White}(x)}{P_{Black}(x)}\right) = \beta^T X$$

(logistic regression)

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$$\min_{\beta} \left\{ -\frac{1}{m} \sum_{i=1}^{m} \left[ y_i \log(h_{\beta}(x_i)) + (1 - y_i) \log(1 - h_{\beta}(x_i)) \right] + \frac{\lambda}{2m} \sum_{j=1}^{n} (\alpha |\beta_j| + (1 - \alpha)\beta_j^2) \right\}$$

## (logistic regression cost function with elastic net regularization)

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$$\min_{\beta} \left\{ -\frac{1}{m} \sum_{i=1}^{m} \left[ y_i \log \left( h_{\beta}(x_i) \right) \right] \right\}$$

(logistic with el



$$\left. - \sum_{j=1}^{n} \left( \alpha \left| \beta_{j} \right| + (1 - \alpha) \beta_{j}^{2} \right) \right\}$$

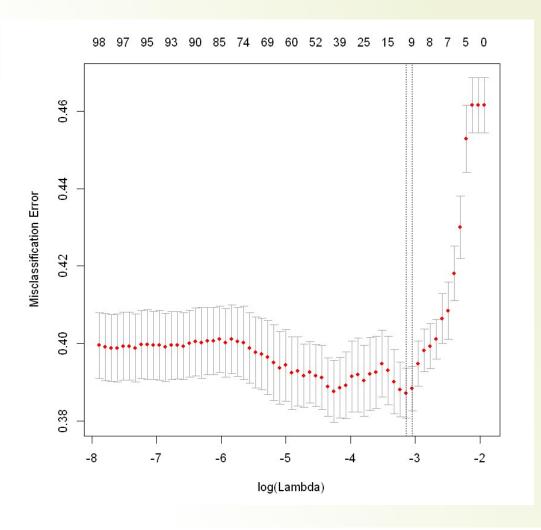
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$$\min_{\beta} \left\{ -\frac{1}{m} \sum_{i=1}^{m} \left[ y_i \log(h_{\beta}(x_i)) + (1-y_i) \log(1-h_{\beta}(x_i)) \right] + \frac{\lambda}{2m} \sum_{j=1}^{n} \left( \alpha |\beta_j| + (1-\alpha)\beta_j^2 \right) \right\}$$

## (logistic regression cost function with elastic net regularization)

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name	coefficient	odds ratio
(Intercept)	-0.34	0.71
Age	0.01	1.01
Feelings of safety: I don't often feel safe.	-0.02	0.98
Feelings of safety: I usually feel safe.	0.12	1.13
Household AMI <10%	0.15	1.16
Household AMI 30-50%	-0.14	0.87
Disabled: No	-0.10	0.91
Disabled: Yes	0.08	1.09
Medical conditions in household: 0	-0.22	0.80
Medical conditions in household: 4+	0.28	1.32

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### What is the definition of vulnerability?

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### From p-values to predictors

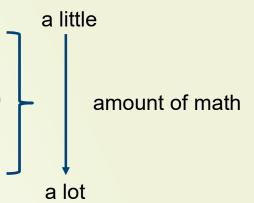
- Books
  - An Introduction to Statistical Learning with Applications in R
  - free Machine Learning courses online (Ng)
  - The Elements of Statistical Learning
  - Deep Learning (deeplearningbook.org)
- Computing languages
  - R, Python, Octave, etc.
- Packages
  - R: glmnet; Python: scikit-learn

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### Thank you!

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