



# Internal benchmarking using propensity scores for detecting racial bias in police traffic stops

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November 3, 2006

# Internal benchmark

Internal  
benchmarking

❖ Internal  
benchmark

❖ Internal  
benchmark

❖ Propensity score  
weighting

Flagging officers

Conclusions

- Consider a particular officer #534
- 71% of this officer's stops involve a black driver

		Percentage
Time	(12-4pm]	9
	(4-8pm]	57
	(8pm-12am]	34
Day	Mon	20
	Tue	12
	Wed	12
	⋮	⋮
Month	Jan	12
	Feb	14
	Mar	7
	Apr	6
	May	8
	⋮	⋮
Area	J	49
	K	33
	L	5
	M	11

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- 46% of similarly situated stops made by other officers involved black drivers

		Percentage	Comparison
Time	(12-4pm]	9	9
	(4-8pm]	57	56
	(8pm-12am]	34	35
Day	Mon	20	20
	Tue	12	11
	Wed	12	12
	⋮	⋮	⋮
Month	Jan	12	12
	Feb	14	15
	Mar	7	7
	Apr	6	6
	May	8	7
	⋮	⋮	⋮
Area	J	49	48
	K	33	34
	L	5	5
	M	11	11

# Propensity score weighting

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- Reweight stops that other officers made so that they have the same distribution of features

$$f(\mathbf{x}|t = 1) = w(\mathbf{x})f(\mathbf{x}|t = 0)$$

- Solving for  $w(\mathbf{x})$  yields the propensity score weight

$$w(\mathbf{x}) = \frac{f(t = 1|\mathbf{x})}{f(t = 0|\mathbf{x})}K = \frac{p(\mathbf{x})}{1 - p(\mathbf{x})}K$$

where  $p(\mathbf{x})$  is the probability that a stop with features  $\mathbf{x}$  involves the officer in question

- Estimate  $p(\mathbf{x})$  using a flexible, non-parametric version of logistic regression
- Compare the percentage of black drivers among the officer's stops with the weighted percentage of black drivers among other stops using weights

$$w_i = p(\mathbf{x}_i)/(1 - p(\mathbf{x}_i))$$

# Results

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❖ Results

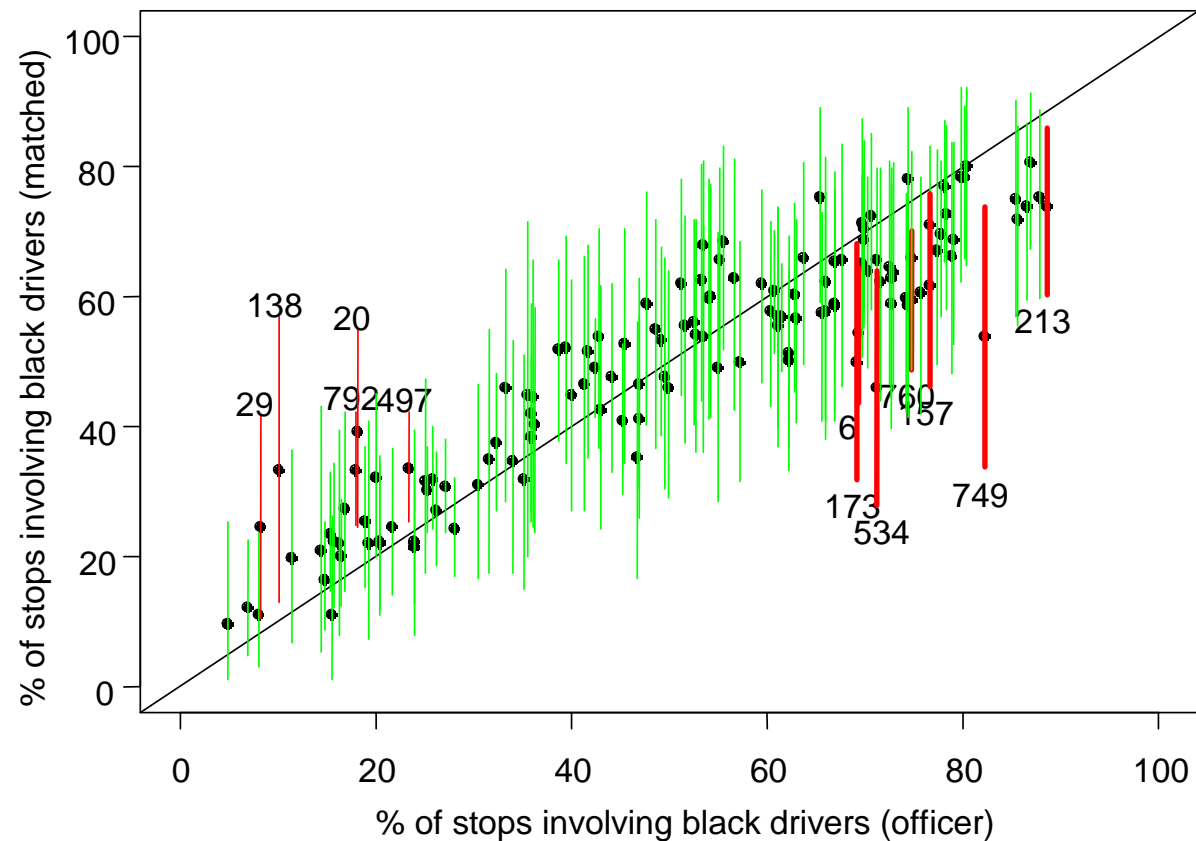
❖ Common  
approach

❖ False discovery  
rate

❖ Estimating fdr

Conclusions

- Seven officers have a substantially greater fraction of stopped black drivers than their internal benchmark



# Common approach

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❖ Results

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approach

❖ False discovery  
rate

❖ Estimating fdr

Conclusions

- A common approach is to compute z-statistics for each officer

$$z = \frac{p_t - p_c}{\sqrt{\frac{p_t(1-p_t)}{n_t} + \frac{p_c(1-p_c)}{ESS}}}$$

- In the absence of racial bias this would be distributed  $N(0,1)$  and a cutoff of 2.0 would be reasonable
- With 133 officers and 133 correlated  $z$ s an appropriate reference distribution can be much wider (Efron 2006).

# False discovery rate

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approach

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rate

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Conclusions

- Benjamini and Hochberg (1995) pioneered the use of the false discovery rate (fdr)

$$\begin{aligned}P(\text{problem}|z) &= 1 - P(\text{no problem}|z) \\&= 1 - \frac{f(z|\text{no problem})f(\text{no problem})}{f(z)} \\&\geq 1 - \frac{f_0(z)}{f(z)}\end{aligned}$$

- If the fraction of problem officers is small then the last inequality is a tight bound



# Estimating fdr

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benchmarking

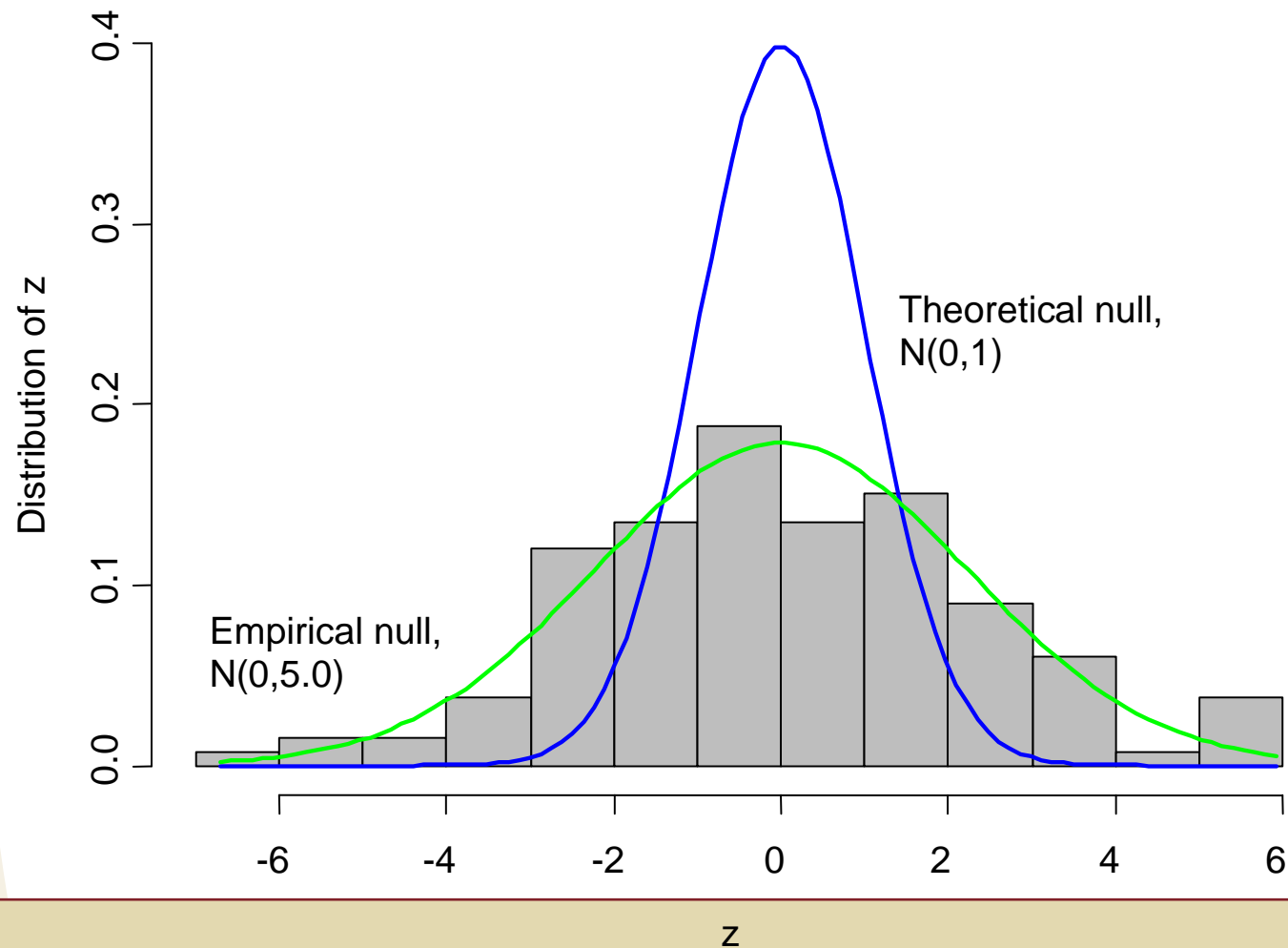
Flagging officers

❖ Results  
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Conclusions

- Estimate  $f_0(z)$  and  $f(z)$  from the observed  $z$ s
- Right tail consists of 5 officers with “problem officer” probabilities ranging from 70% to 86%



# Conclusions

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Conclusions

❖ Conclusions

- Internal benchmarking can help identify problem officers
- Propensity score weighting offers a sound process for constructing the internal benchmark
- Flagging particular officers requires dealing with the issues of massive multiple comparisons
- False discovery rate offer a promising direction