

Using Machine Learning Forecasts to Improve Evaluation

Geoffrey C. Barnes, Ph.D. Western Australia Police University of Cambridge

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DURHAM CHECKPOINT EVALUATION

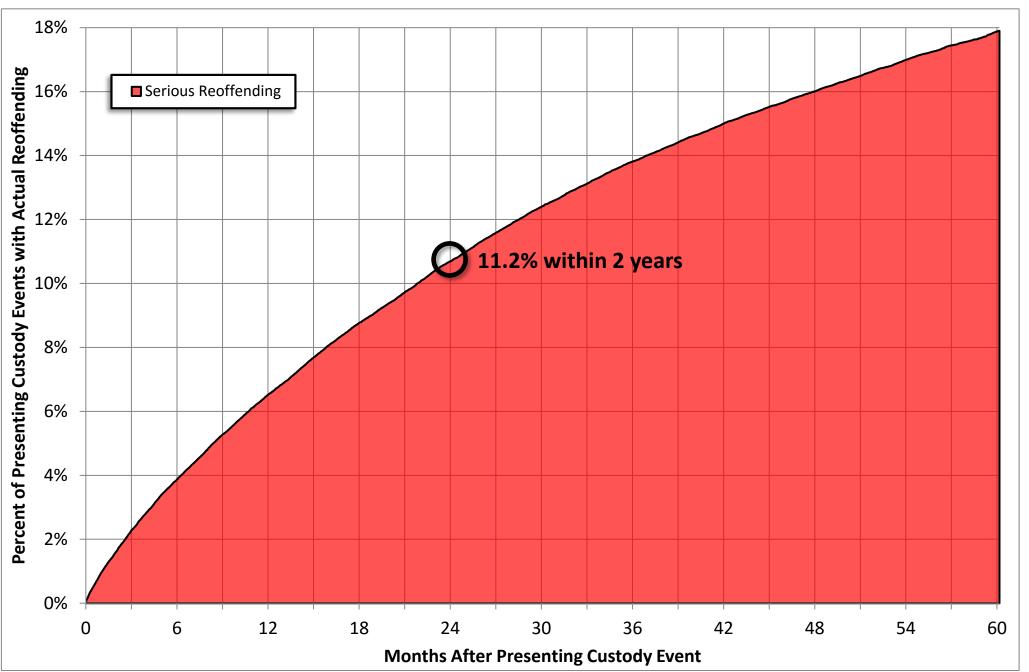
What Outcomes Are Predicted by HART (Harm Assessment Risk Tool)?

Three different outcome categories

- Combines the prevalence of offending (i.e., yes/no) with the seriousness
- <u>High Risk</u>: A new serious offence occurring within 2 years

Murder Attempted murder Aggravated violence (i.e., GBH) Robbery Any sexual offence Any firearm offence







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Murder

Attempted murder

Aggravated violence (i.e., GBH)

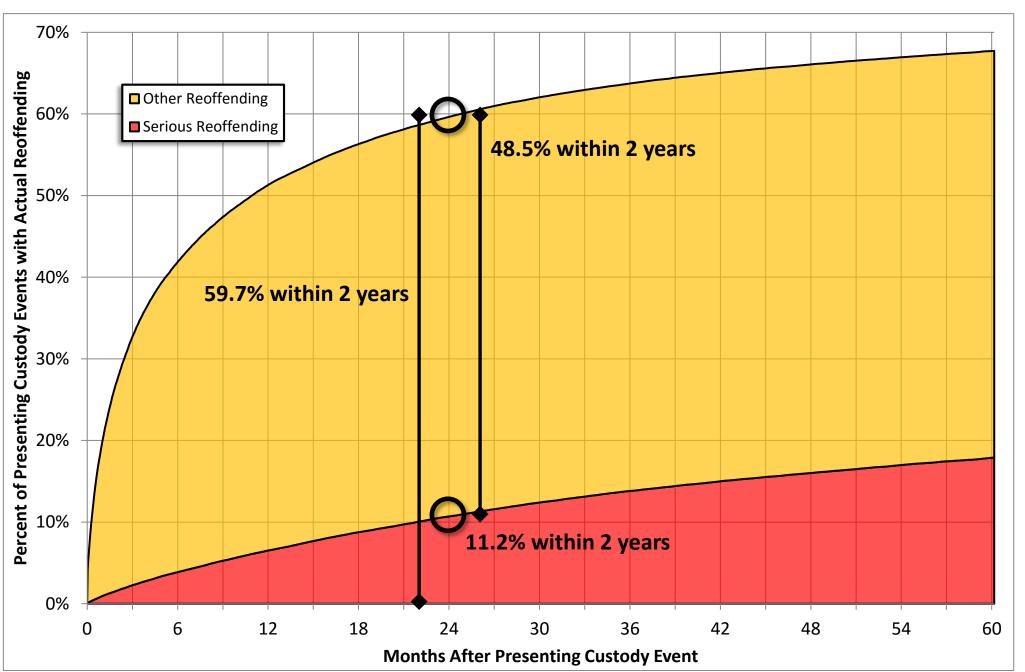
Robbery

Any sexual offence

Any firearm offence

• Moderate Risk: Any new non-serious offence within 2 years







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Aggravated violence (i.e., GBH)

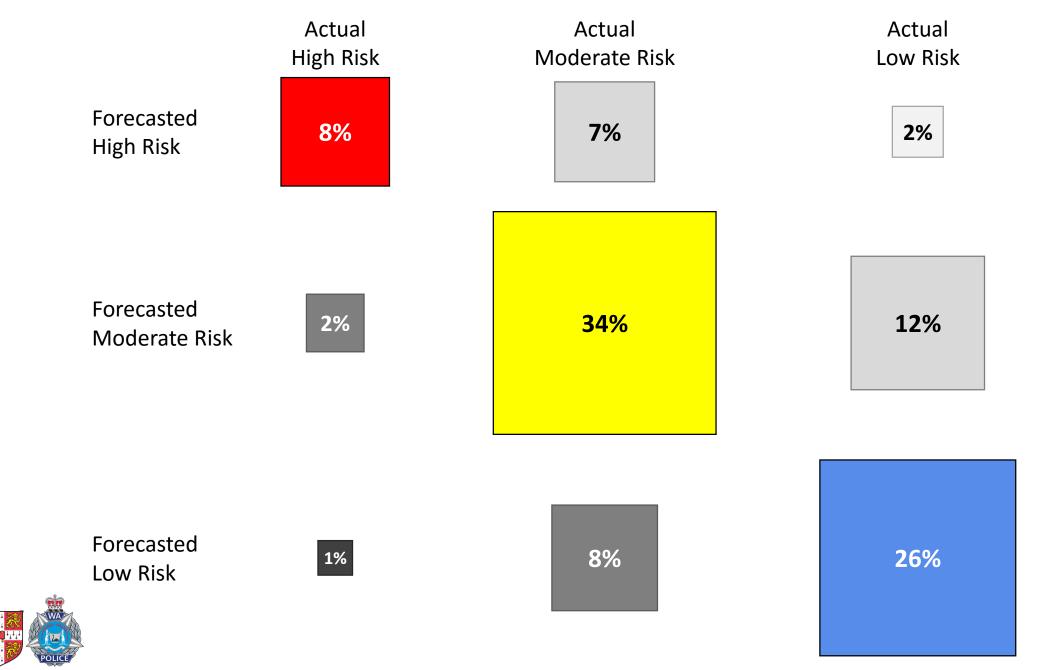
Robbery

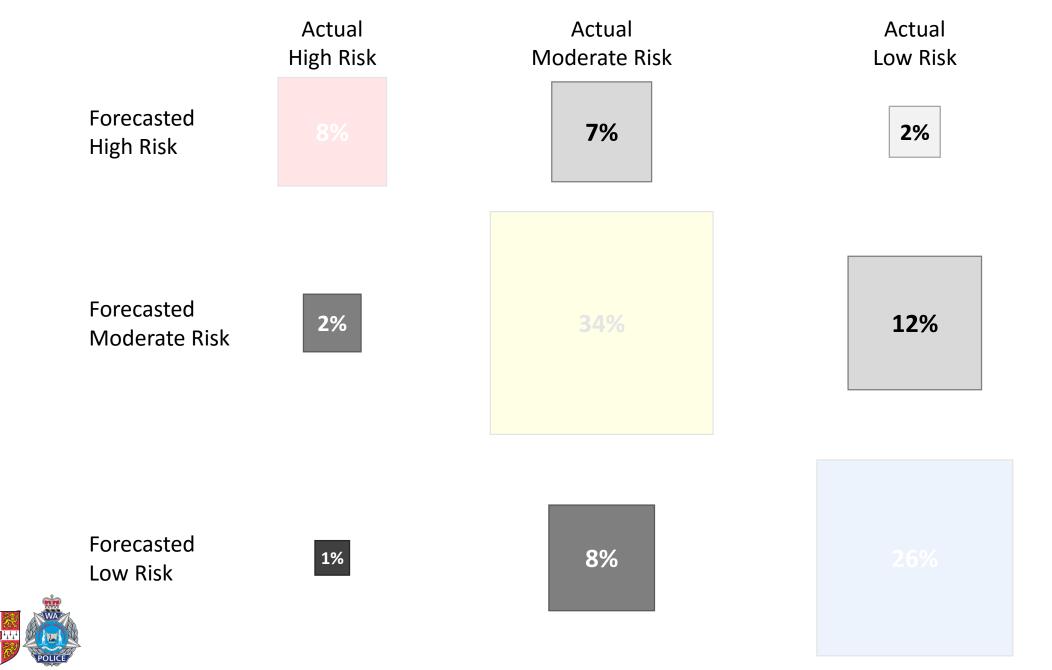
Any sexual offence

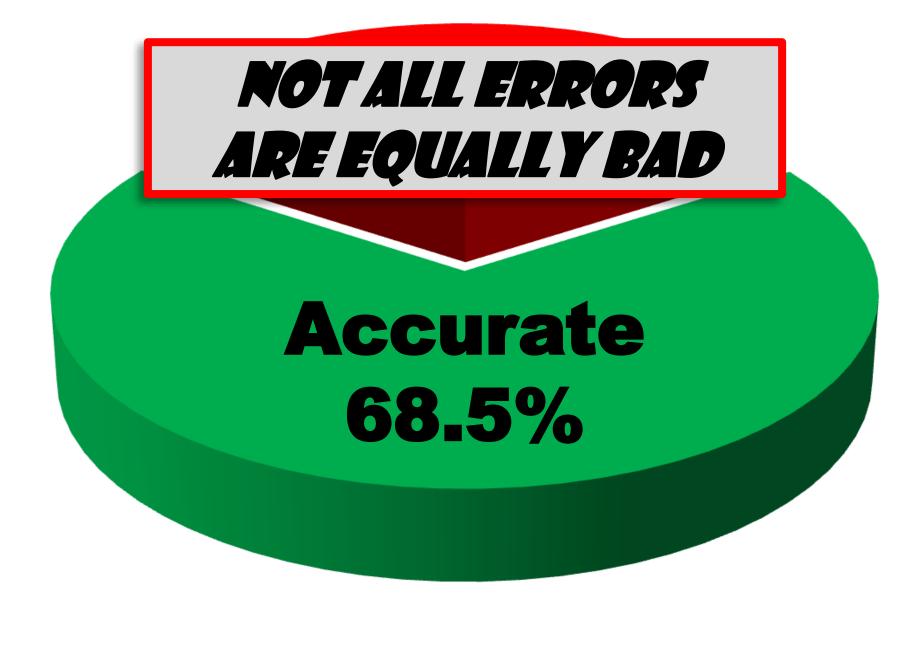
Any firearm offence

- Moderate Risk: Any new non-serious offence within 2 years
- Low Risk: No new offences of any kind within 2 years

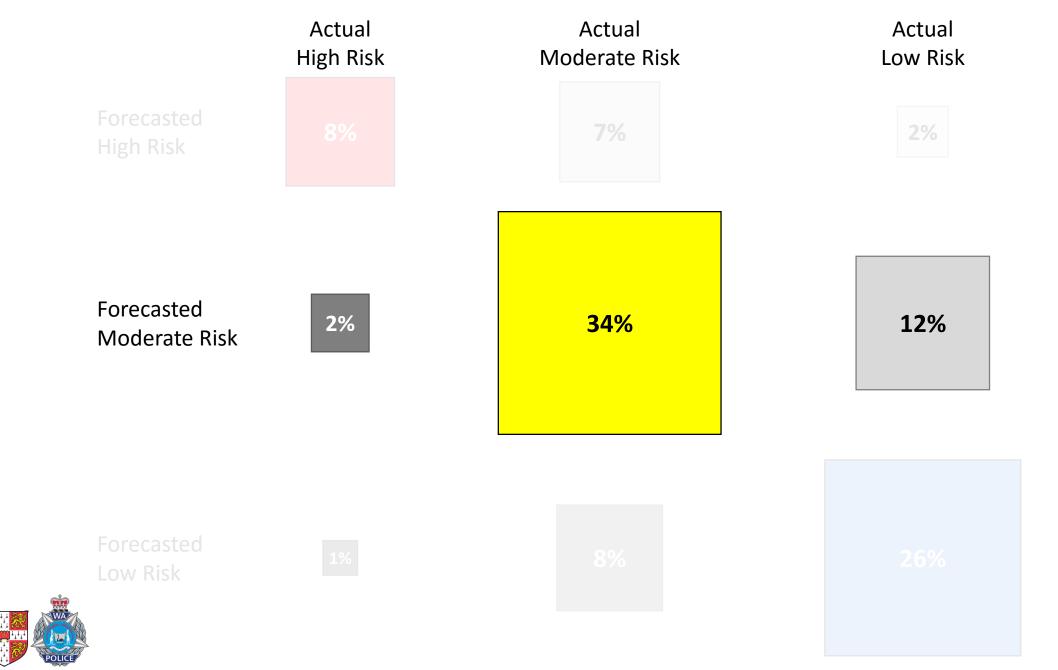




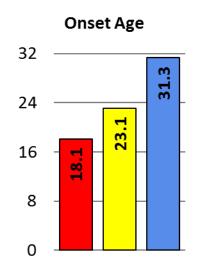




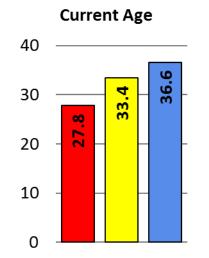


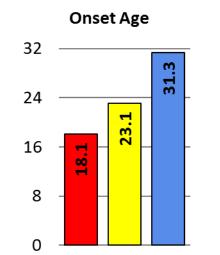


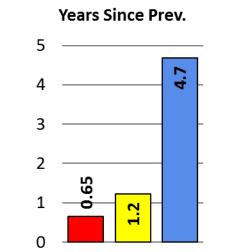


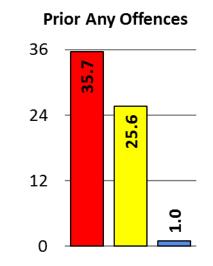




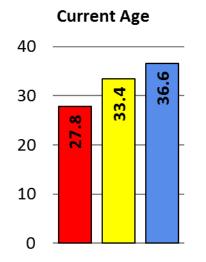


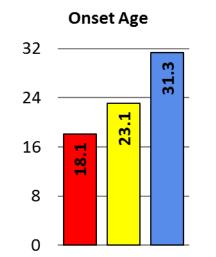


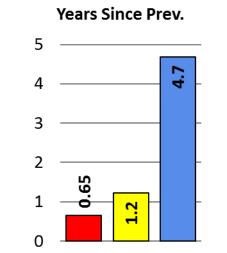


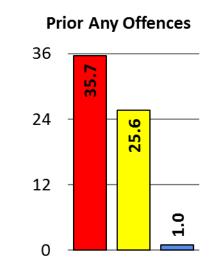




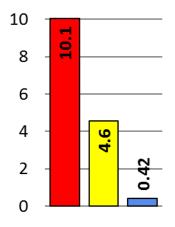




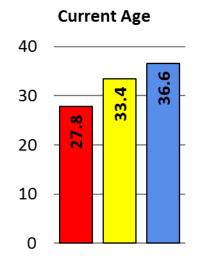


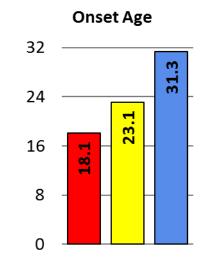


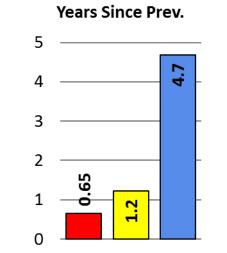
Prior Violence

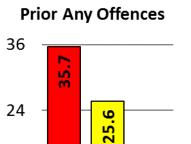










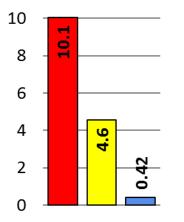


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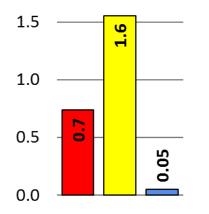
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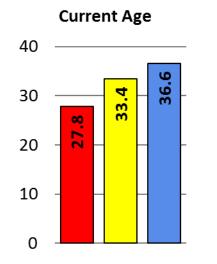
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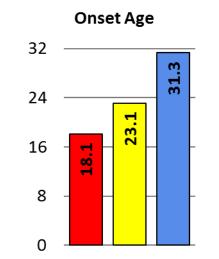


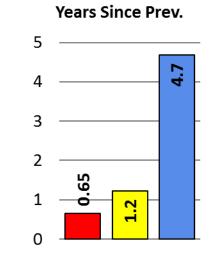
Prior Drug

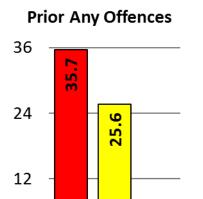






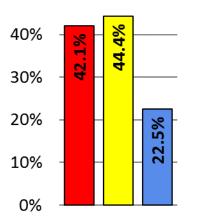






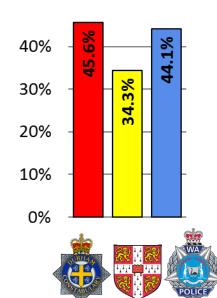
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Instant Property

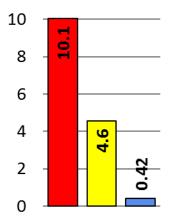


Instant Violence

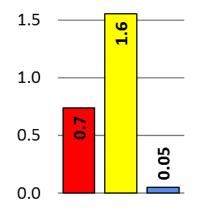
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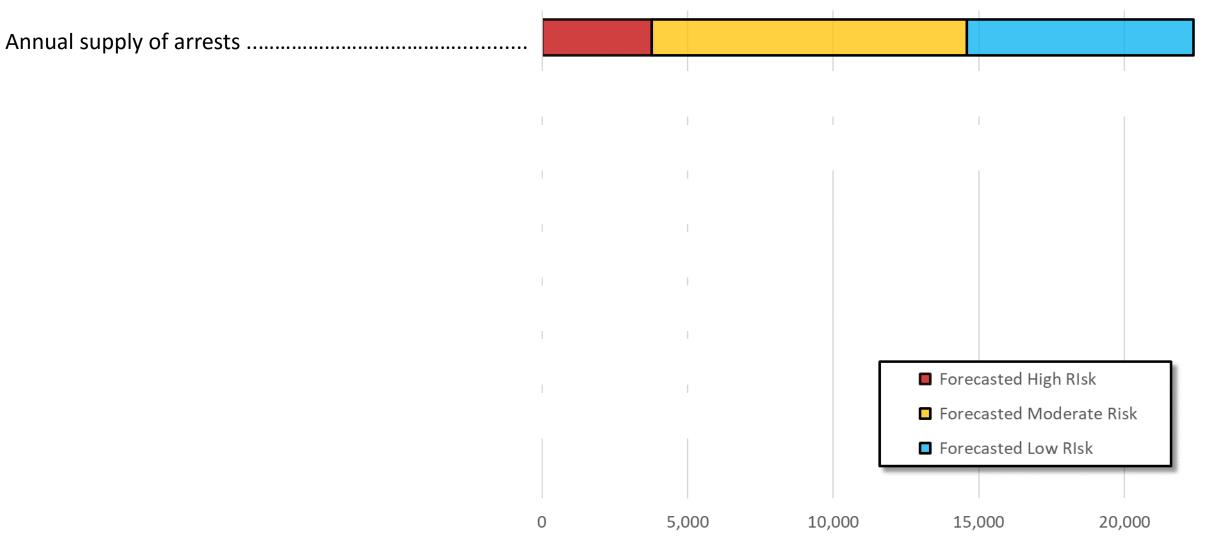
Prior Violence



Prior Drug

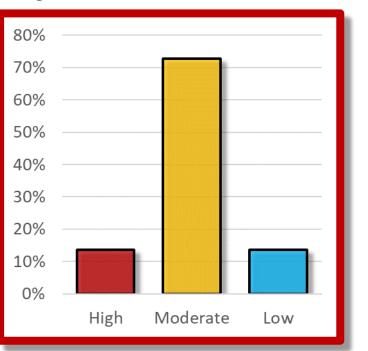


Typical Enrollment Attrition in Criminal Justice Experiments:

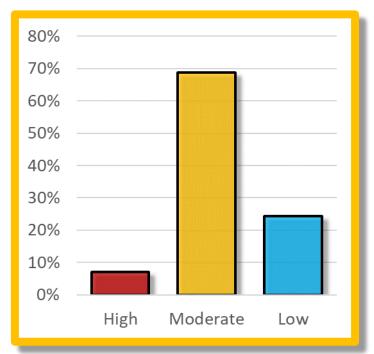


What Do Durham Custody Sergeants Say In Their Own Assessments of Risk?

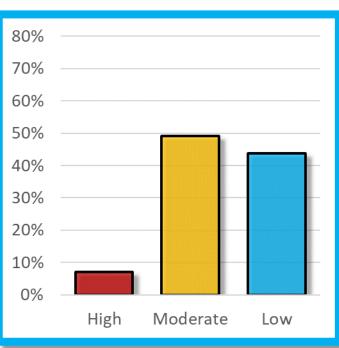
When the model's forecast is High Risk...



When the model's forecast is Moderate Risk...



When the model's forecast is Low Risk...

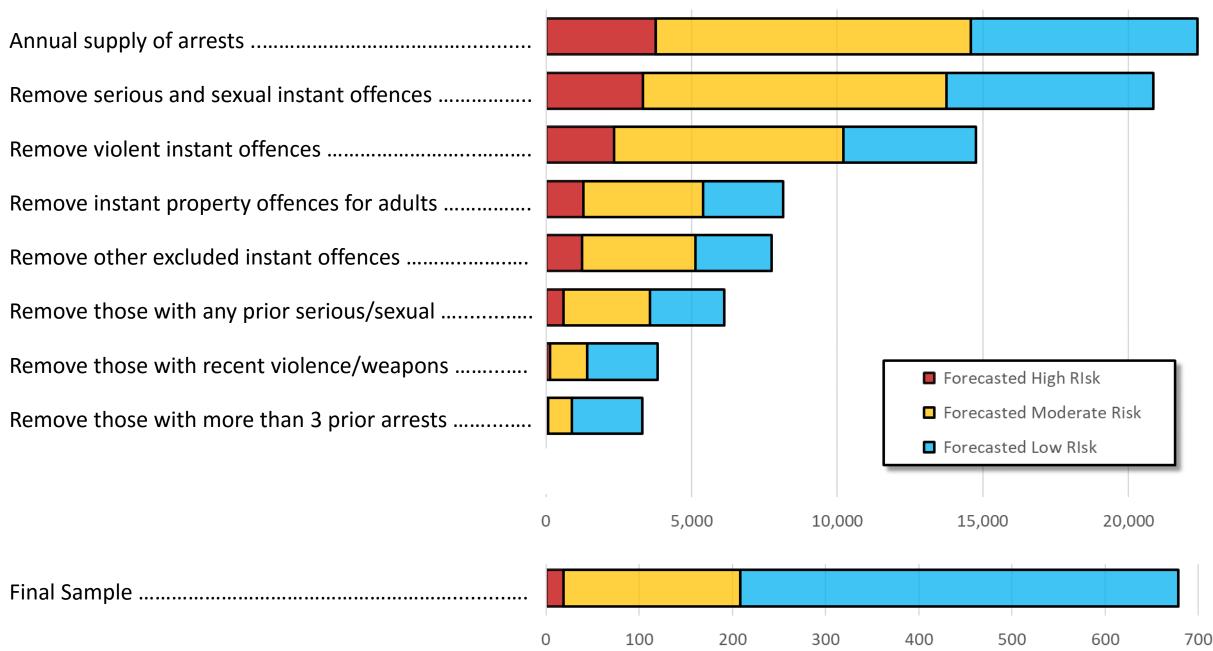


When the Sergeant thinks somebody is High Risk... they are the least likely to make a refer for random assignment (5%)

When the Sergeant thinks somebody is Moderate Risk... they will refer for random assignment when they remember (30%)

When the Sergeant thinks somebody is Low Risk... they will resist referral to avoid the risk of control assignment (10%)

Typical Enrollment Attrition in Criminal Justice Experiments:



So what happens when a random forest is used to govern eligibility?

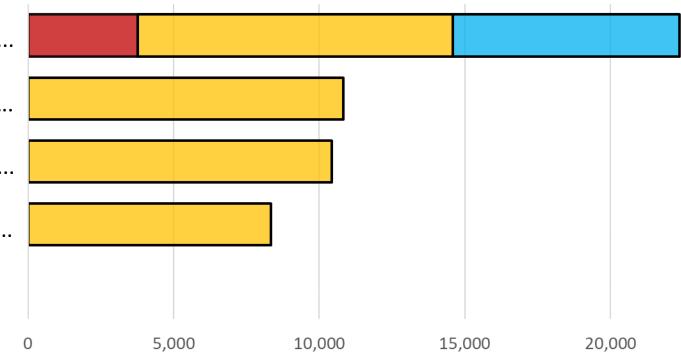
Better Experimental Enrollment Using Random Forest Forecasts:

Annual supply of arrests

Remove forecasted low and high risk cases

Remove serious and sexual instant offences

Remove another 20% for other eligibility rules



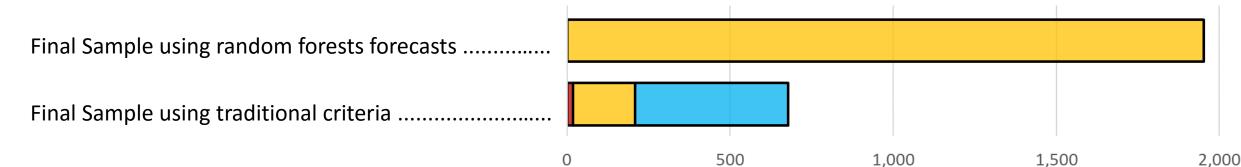
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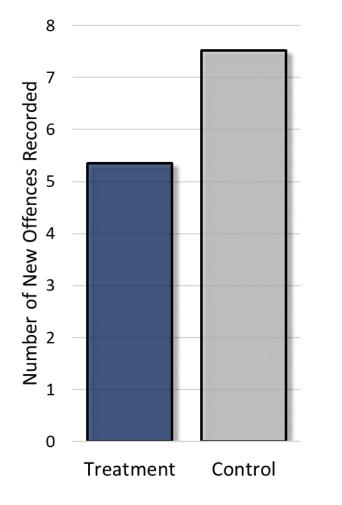
 0
 5,000
 10,000
 15,000
 20,000

Comparison of the two samples:

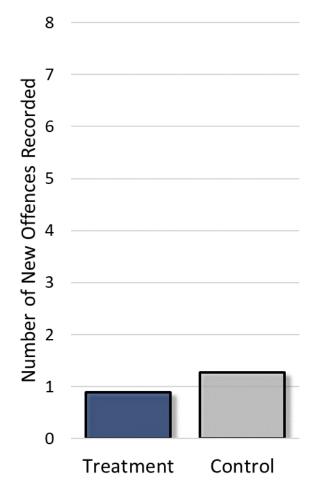


Monte Carlo Simulation of Different Eligibility Rules:

When the random forest model is used to determine eligibility...



When traditional offencebased criteria are used to determine eligibility...



Difference Between Treatment and Control:

Random Forest:	-2.17
Traditional:	-0.38

Percentage Decrease in Offending:

Random Forest:	28.9%
Traditional:	29.9%

Standarised Difference in Means (t):

Random Forest:	7.94
Traditional:	1.84

Average Significance Result (p):

Random Forest:	1.35 x 10 ⁻⁹
Traditional:	0.161

Statistical Power (Probability of *p* < .05):

Random Forest:	100%
Traditional:	30%

Two Things that Cops Hate