Predicting offending on bail

Felix Leung, BOCSAR

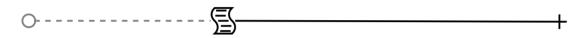
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- 2. Predictive modelling

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- 2. Predictive modelling
- 3. Results

1. Aim

1. Bail dispensed with



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2. Bail granted



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2. Bail granted



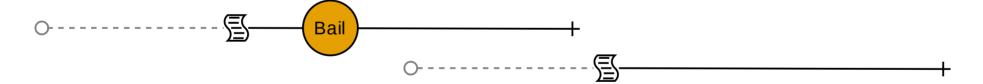
3. Bail refused



1. Bail dispensed with



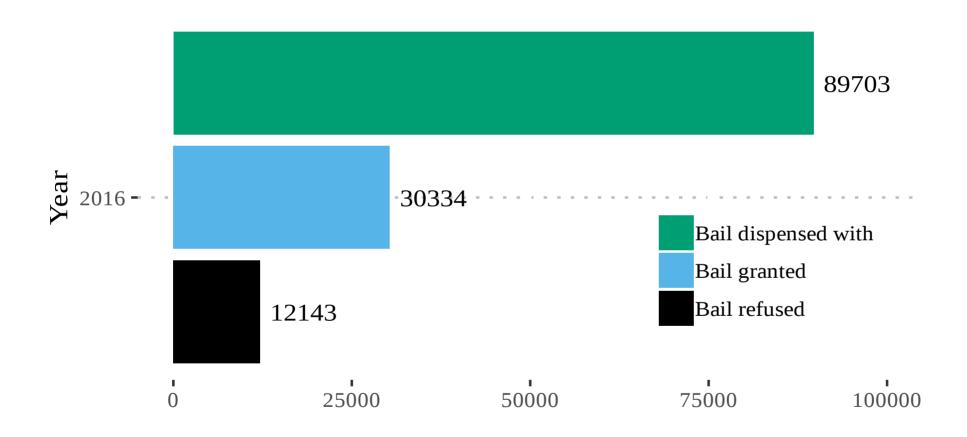
2. Bail granted



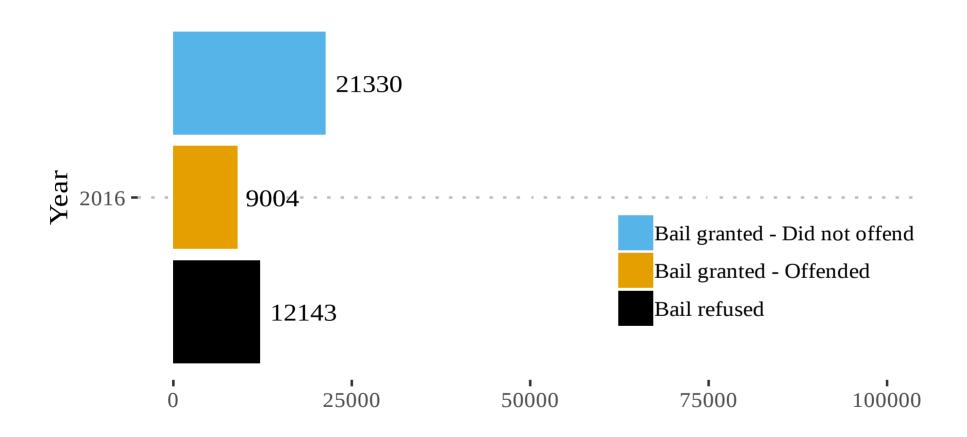
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Bail outcomes in 2016



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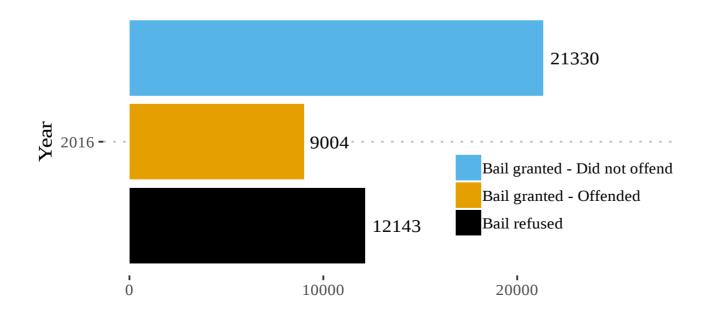
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2. Predictive modelling

Data

NSW Reoffending Database

Data

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- Include:
 - Defendants charged between 2011 and 2016

Data

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- Include:
 - Defendants charged between 2011 and 2016
- Exclude:
 - ANZSOC subdivision 152 Breach of community-based order
 - ANZSOC group 1569 Offences against justice procedures, nec

(ANZSOC: Australian and New Zealand Standard Offence Classification)

• Outcome: new proven offence committed on bail

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- **Test data**: 2016

Predictors

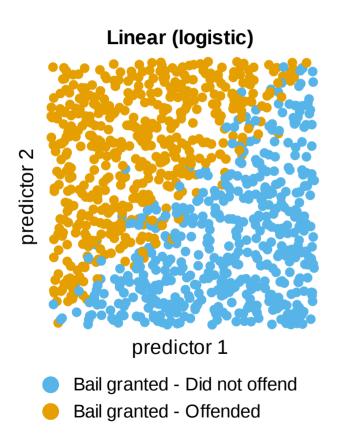
- 1. Demographics:
 - e.g. age, gender, socio-economic index (SEIFA)

Predictors

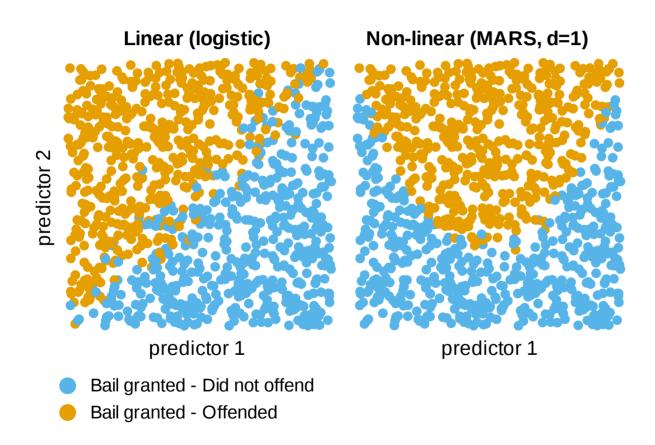
- 1. Demographics:
 - e.g. age, gender, socio-economic index (SEIFA)
- 2. Variables on index offence:
 - 。 e.g. whether offence indictable, summary, DV-related
 - ANZSOC divisions and subdivisions

Predictors

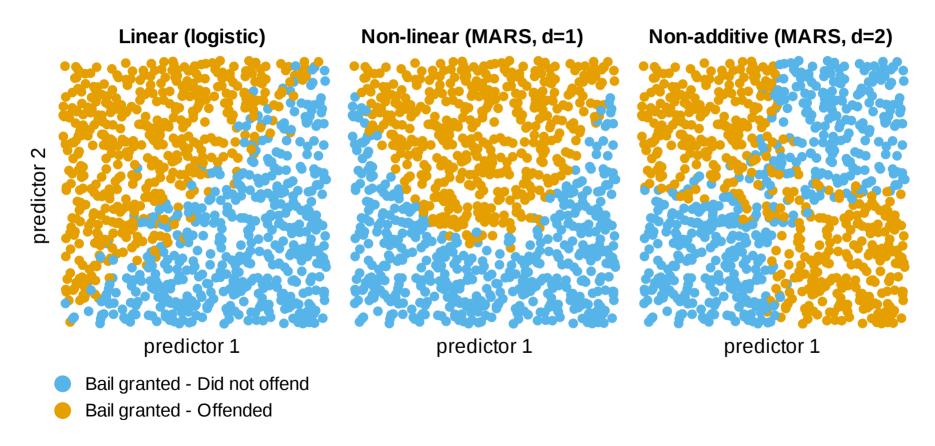
- 1. Demographics:
 - e.g. age, gender, socio-economic index (SEIFA)
- 2. Variables on index offence:
 - e.g. whether offence indictable, summary, DV-related
 - ANZSOC divisions and subdivisions
- 3. Criminal history:
 - Age at first criminal justice contact, no. guilty charges
 - Previous penalty: prison, fine, bond, etc
 - and more



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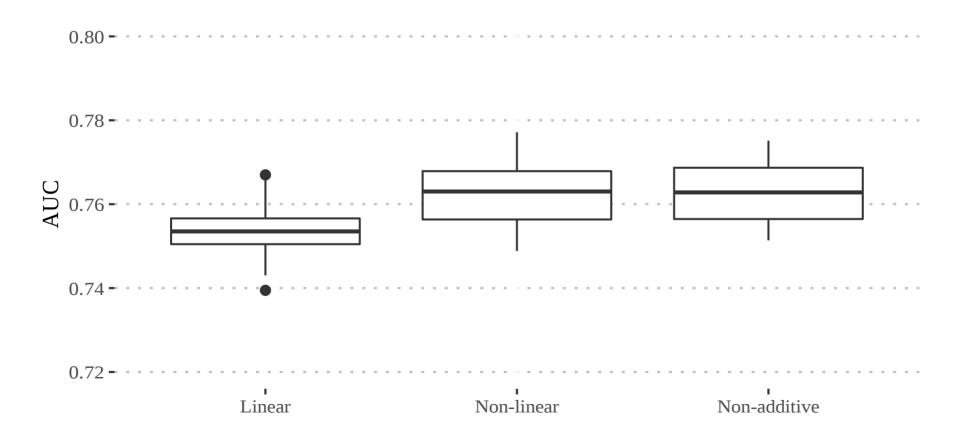
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Model selection

based on AUC (area under curve)

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1. Demographics:

Age at charge date

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2. Variables on index offence:

- Median Severity Ranking of most serious offence charged
- Most serious offence in ANZSOC 03 (Sexual assault)
- Most serious offence in ANZSOC 08 (Theft)
- Most serious offence in ANZSOC 151 (Breach custodial order)

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Age at charge date

2. Variables on index offence:

- Median Severity Ranking of most serious offence charged
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- Most serious offence in ANZSOC 151 (Breach custodial order)

3. Criminal history:

- Age at first criminal justice contact
- No. of prev. charges
- No. of prev. prison
- o Proportion of prev. bail with new proven offence
- Time since prev. charge

3. Results

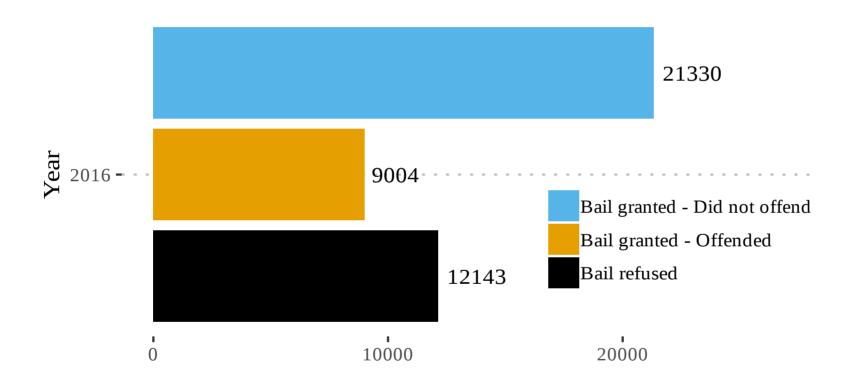
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- 2. To compare model predictions with 2016 figures, some threshold is needed
- 3. Here we assume *hypothetically*:
 - o individual above the threshold would be refused bail
 - o individual **below** the threshold would be granted bail
- 4. Method originated from Kleinberg, Lakkaraju, Leskovec, Ludwig, and Mullainathan (2017)

Bail outcomes in 2016



Model Predictions

21330

Bail granted - Did not offend

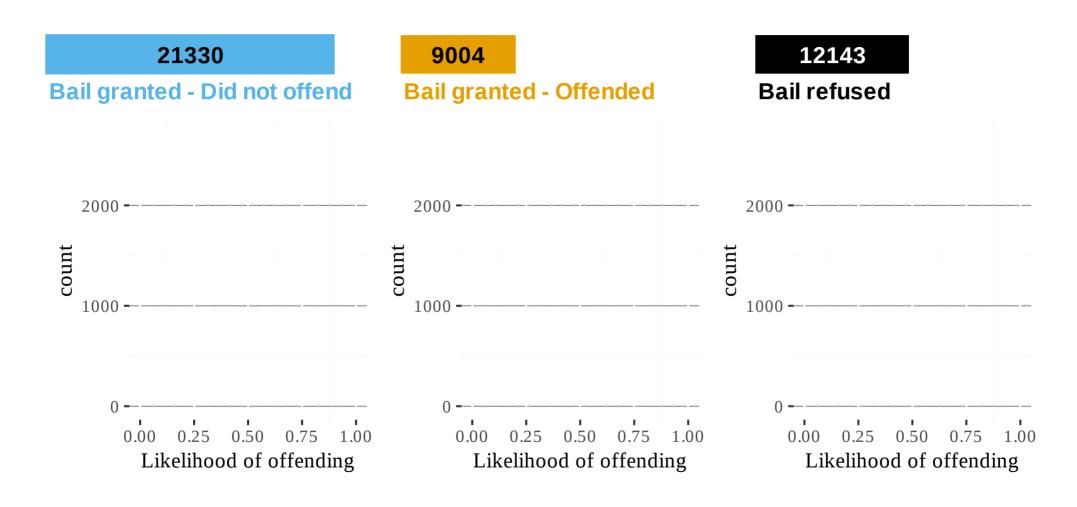
9004

Bail granted - Offended

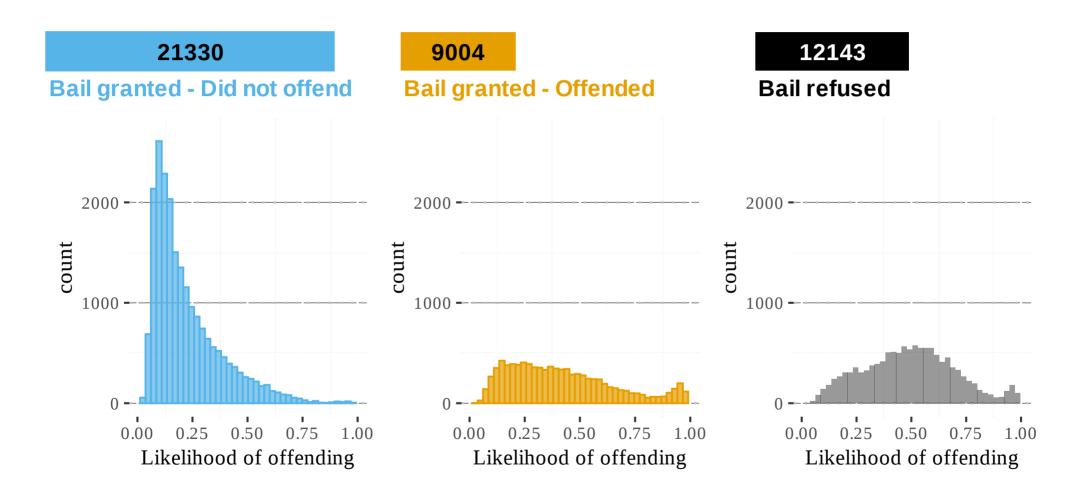
12143

Bail refused

Model Predictions



Model Predictions



Aim

Could we:

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- 2. Reduce bail refusals (without increasing offending on bail)?

21330

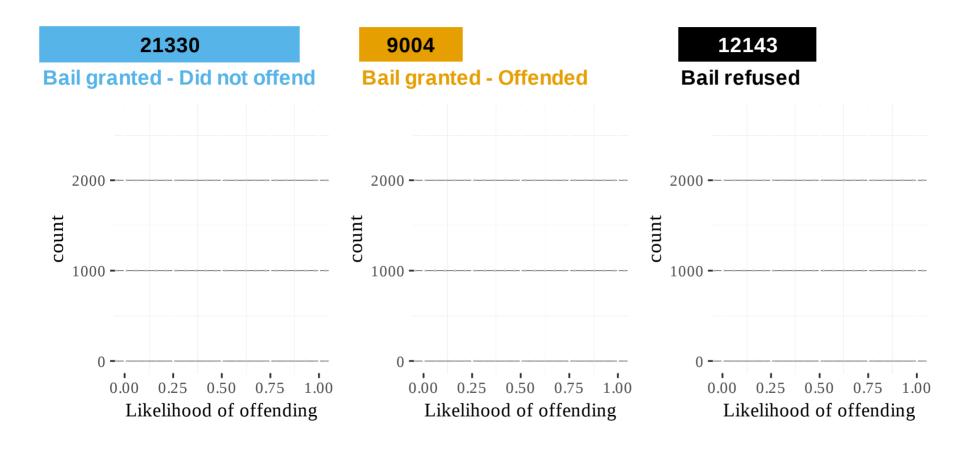
Bail granted - Did not offend

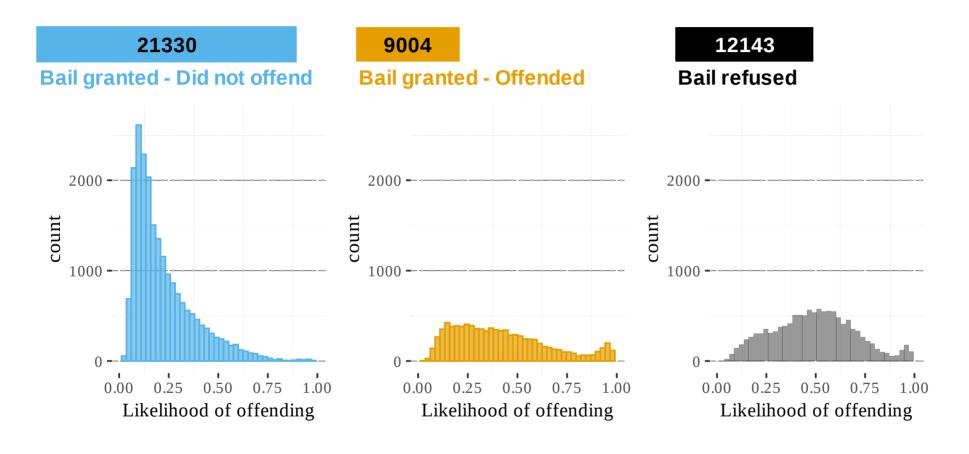
9004

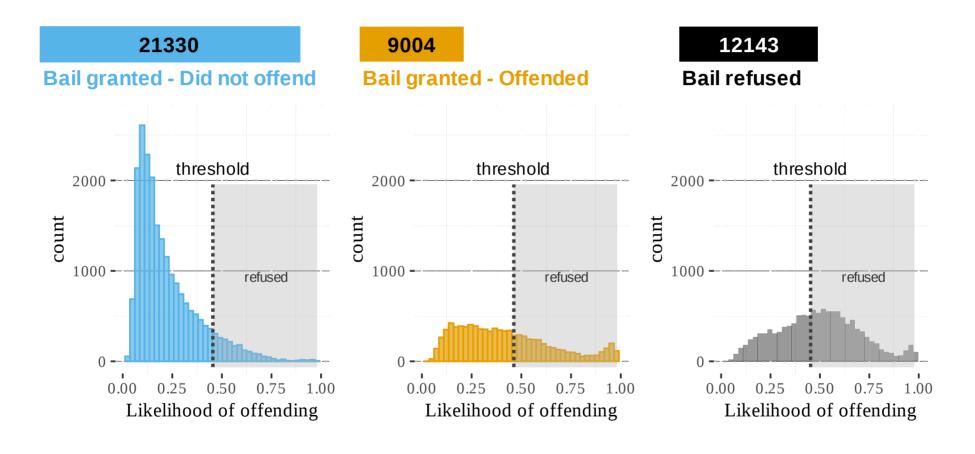
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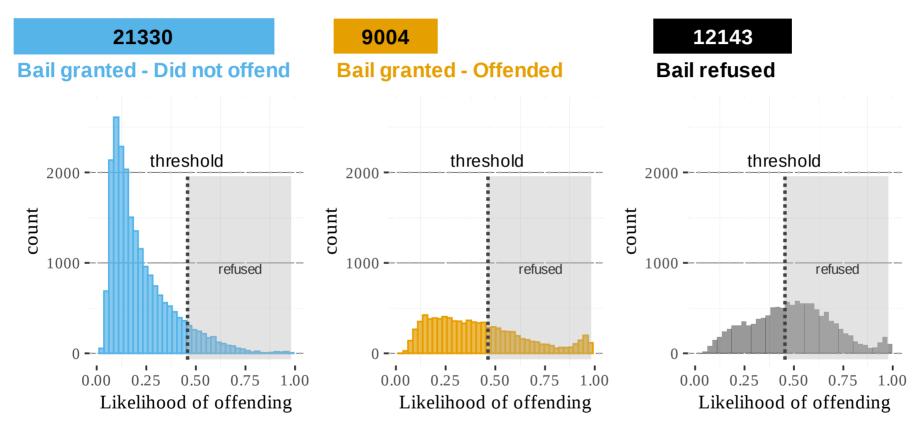
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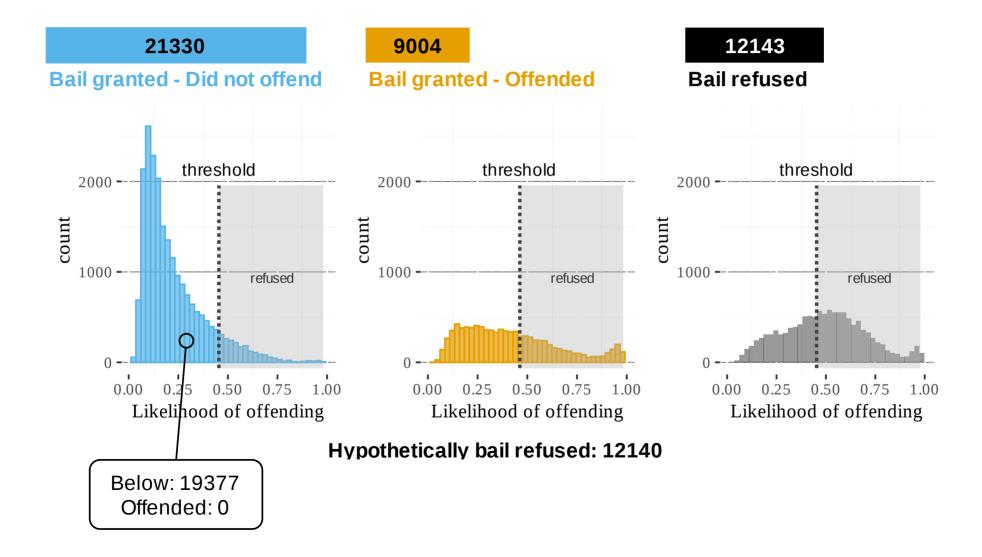


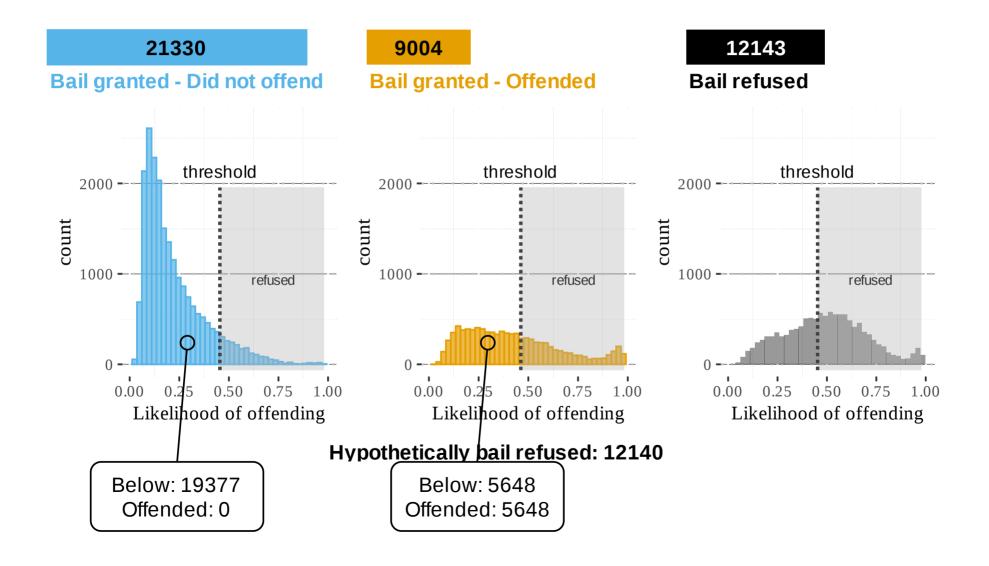


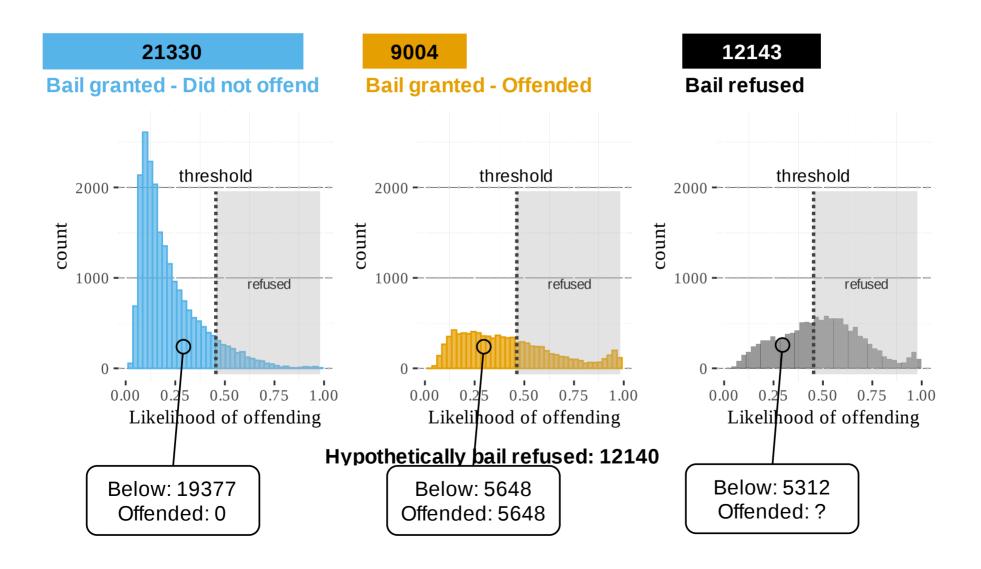


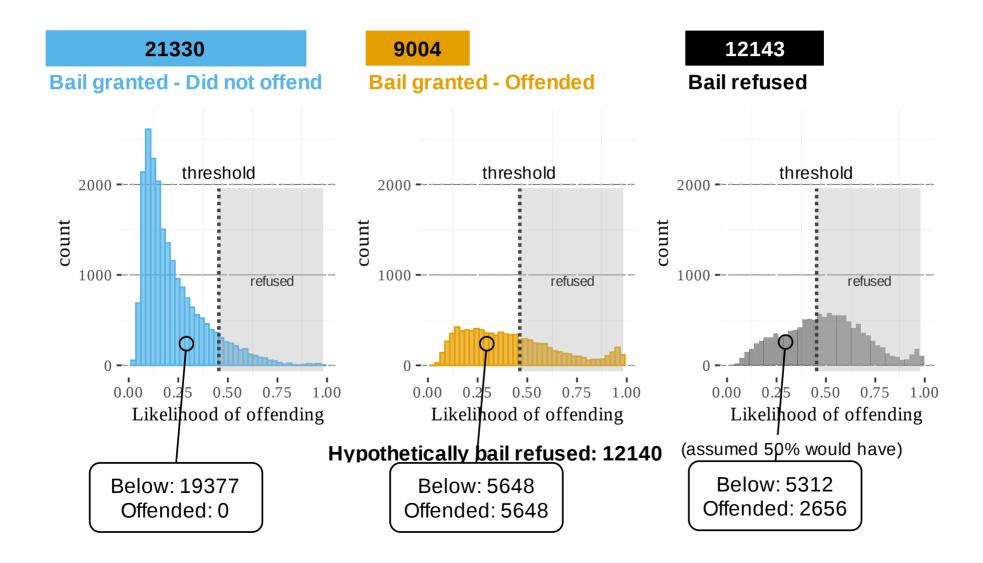


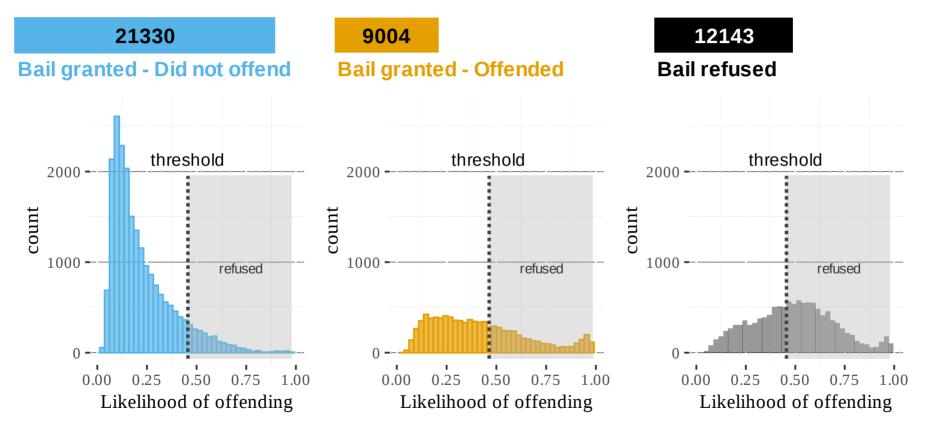
Hypothetically bail refused: 12140





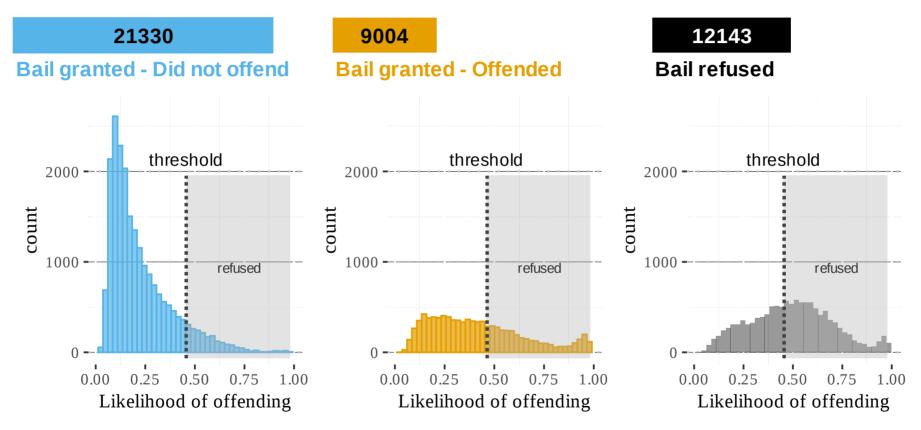






Hypothetically bail refused: 12140

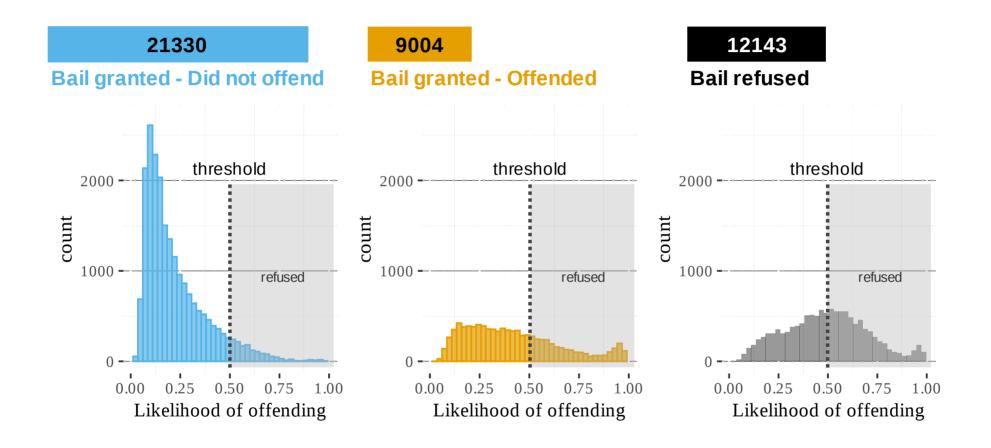
Hypothetically bail granted - offended: 5648 + 2656 = 8304

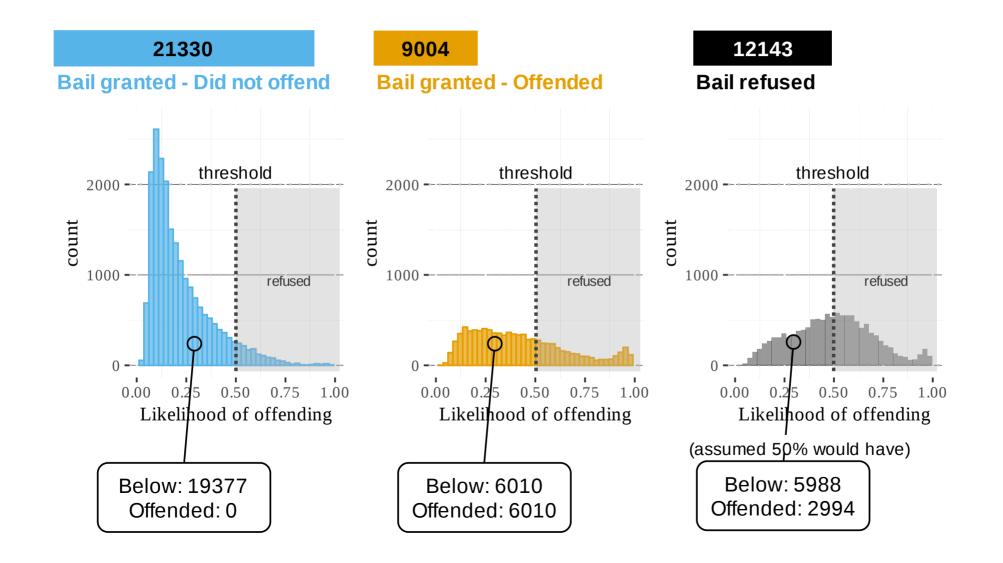


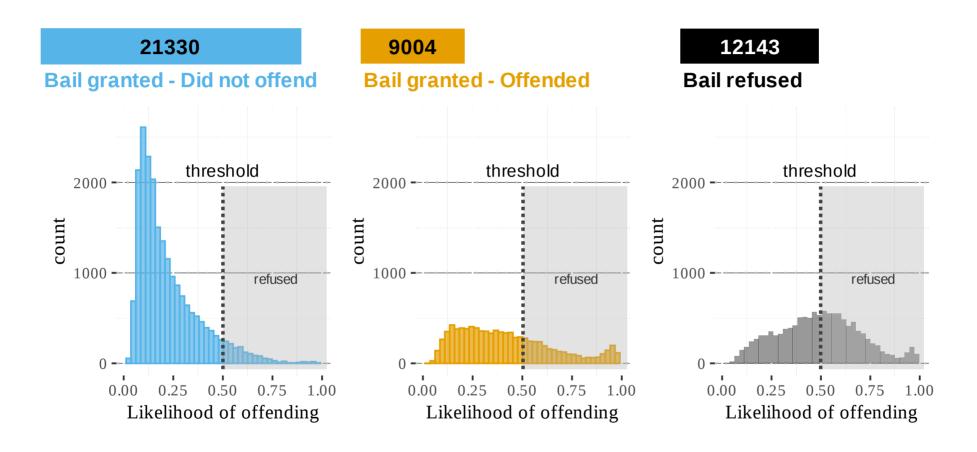
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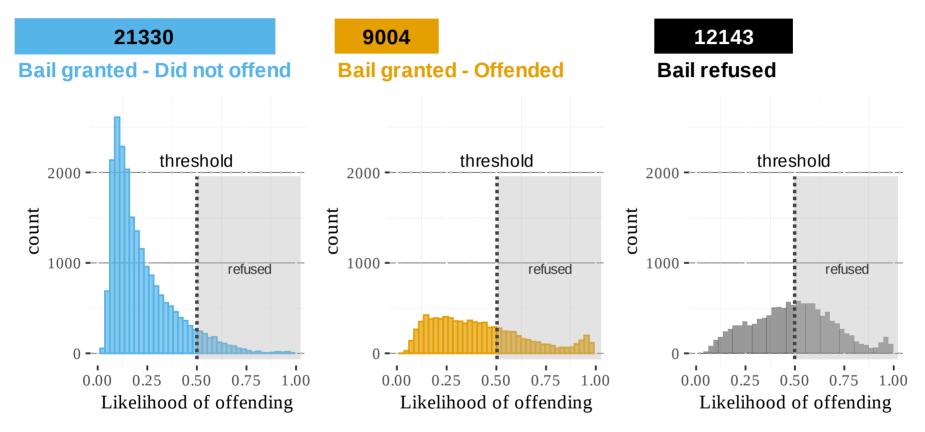
(a reduction of 700 or 7.8%)





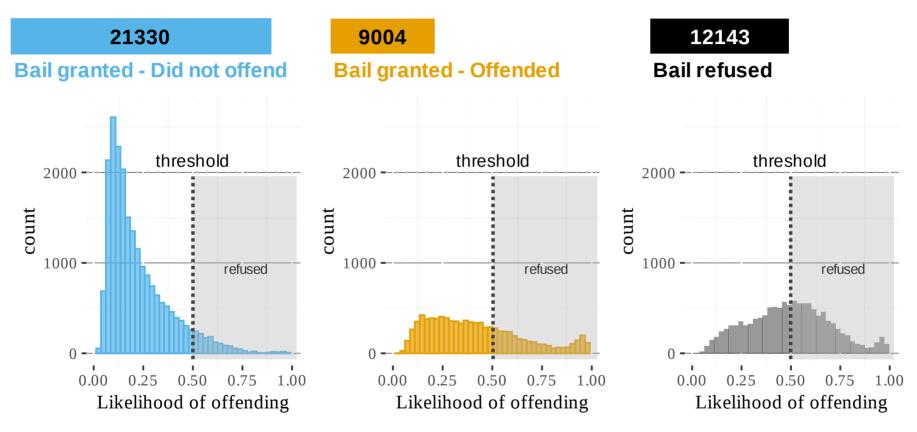


Hypothetically bail granted - offended: 6010 + 2994 = 9004



Hypothetically bail refused: 10750

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(a reduction of 1393 or 11.5%)

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What we found:

- 1. offending on bail: 700 or 7.8% reduction
- 2. bail refusals: 1393 or 11.5% reduction

(this is assuming 50% of those bail refused but below the threshold would have offended if granted bail.)

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- 2. Offending is measured crudely; some of the offending predicted would not be serious enough to change a bail decision
- 3. Other important concerns such as fairness, non-representative data are not addressed

References

- [1] J. H. Friedman. "Multivariate adaptive regression splines". In: *The annals of statistics* (1991), pp. 1-67.
- [2] J. Kleinberg, H. Lakkaraju, J. Leskovec, et al. "Human decisions and machine predictions". In: *The quarterly journal of economics* 133.1 (2017), pp. 237-293.