

Re-sampling, Costsensitive learning and Probability Calibration

### ML model outputs and probability

Logistic Regression returns calibrated probabilities

- Some machine learning models return uncalibrated probabilities
  - Decision trees
  - ➤ Naïve Bayes



### Imbalance data techniques and probabilities

Over-sampling, under-sampling and cost-sensitive learning distort the relationship between the returned probabilities and the fraction of positive observations.

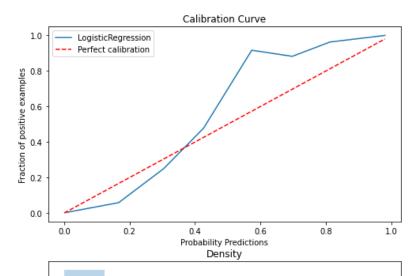
- > The first distort the distribution of classes.
- > The second modifies the learning function.





## Re-sampling – Logistic Regression





0.8

0.6

Probability Predictions

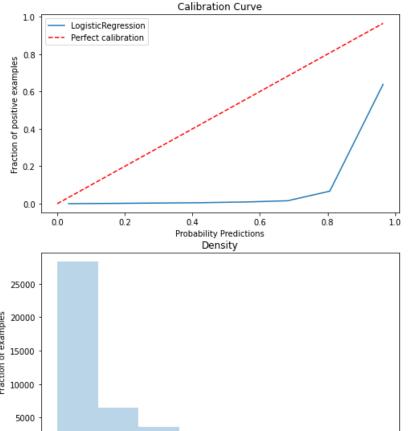
1.0

0.0

0.2

0.4

#### Random under-sampling



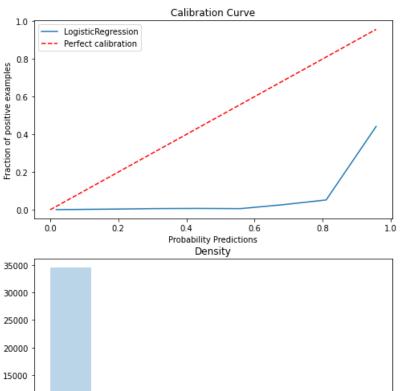
0.6

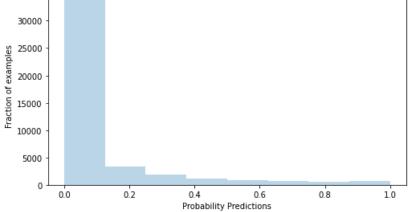
Probability Predictions

0.8

1.0

#### **Borderline SMOTE**







0.0

0.2

40000

30000

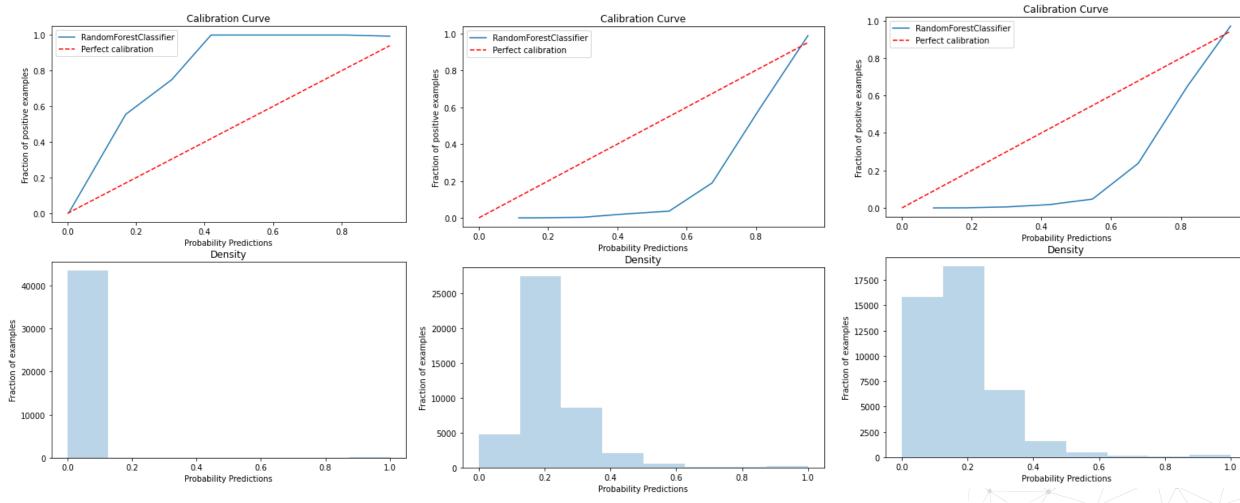
10000

### Re-sampling – Random Forest



### Data Random under-sampling

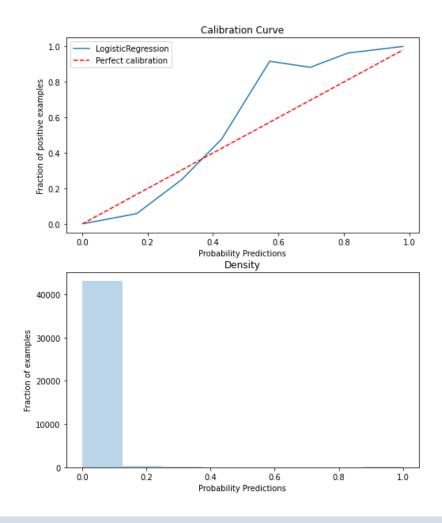
#### mpling Borderline SMOTE



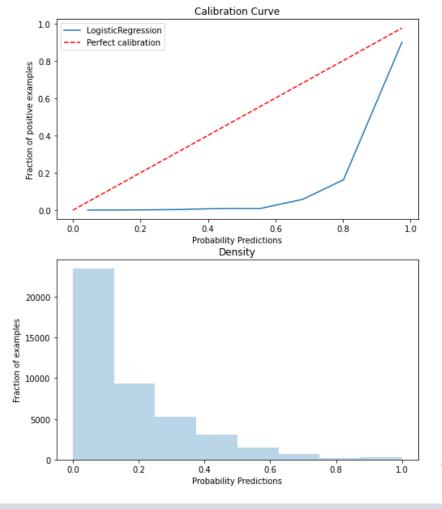


### Cost-sensitive – Logistic Regression

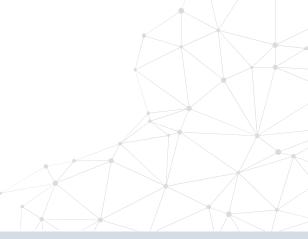
#### Raw Data



#### **Cost-Sensitive Learning**



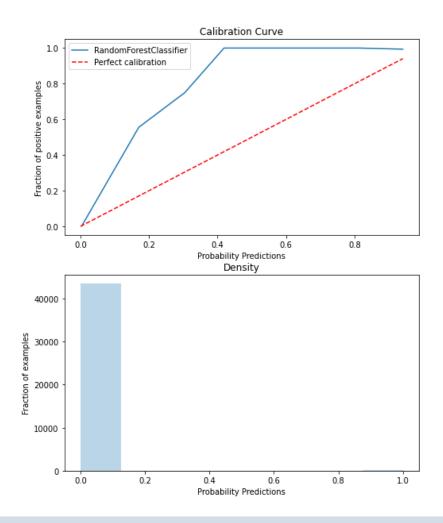
Very similar to Random Undersampling!!!



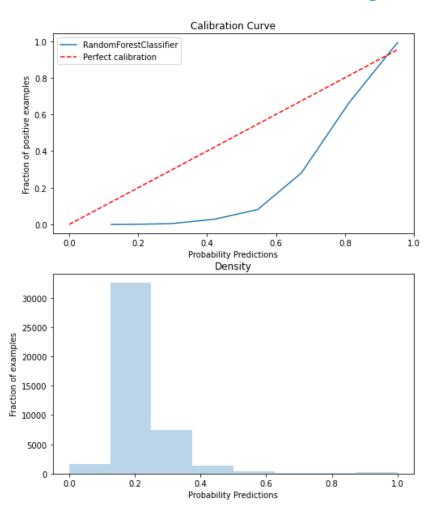


### Cost-sensitive – Random Forest

#### Raw Data



#### **Cost-Sensitive Learning**



Very similar to Random Undersampling!!!





## Probability as certainty

Probabilities can be much more informative than labels.

• "The model predicts this claim is fraudulent" vs "The model predicts this claim is 90% likely to be fraudulent"

• To convey likelihood, we need <u>calibrate</u> the probabilities after resampling or cost-sensitive learning





# THANK YOU

www.trainindata.com