

The idea is to retain observations from the majority that are hard to classify, but remove the noise.

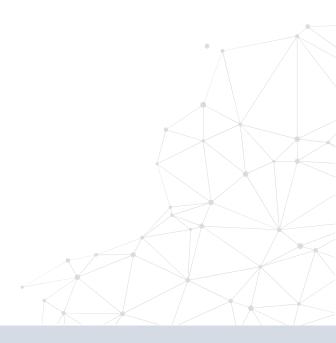




- First, selects samples at the boundary of the classes (hardest instances).
- Next, removes the Tomek Links (noise)



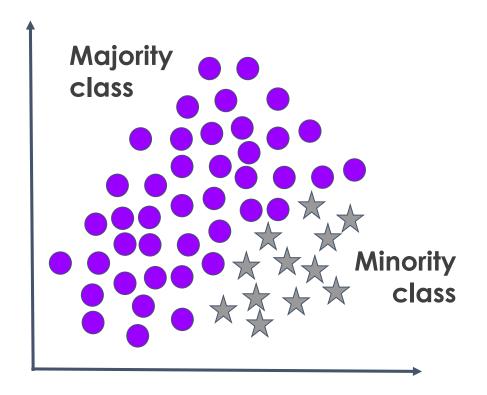
- Cleaning
- Final dataset shape varies
- Boundary matters



#### One Sided Selection: Procedure

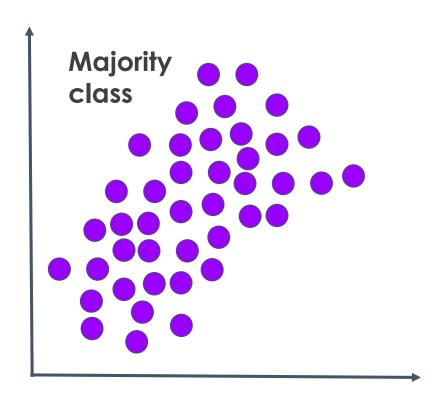
- 1. Create group S with all samples from minority
- 2. Add 1 observation from the majority to S (at random)
- 3. Train a 1-KNN on S
- 4. Make predictions on the rest of the majority class observations
- 5. If predictions don't match the class, pass the samples to \$
- 6. In S, find and and remove Tomek Links









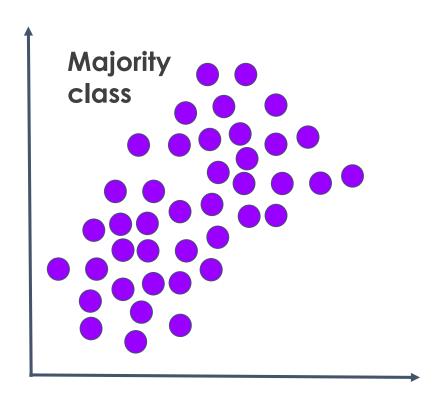


Separate minority class into a group









Take 1 observation from Majority class to Minority group

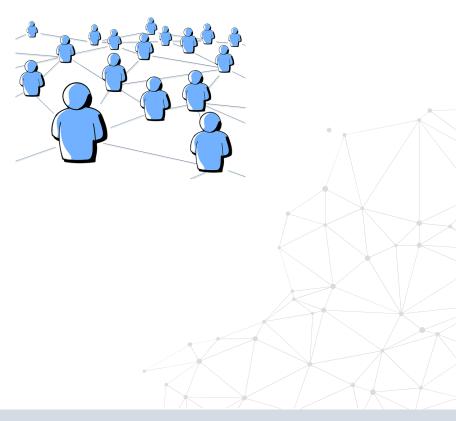


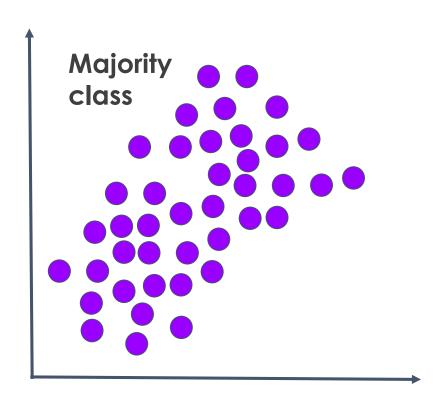




Train a 1 KNN algorithm





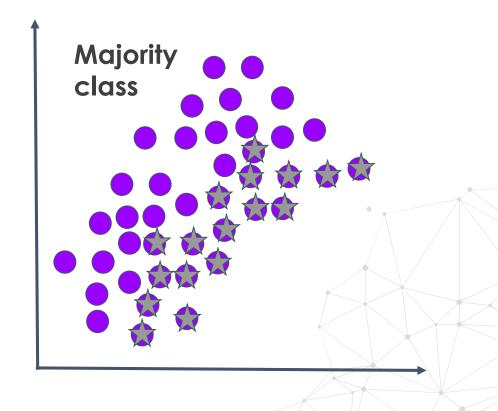




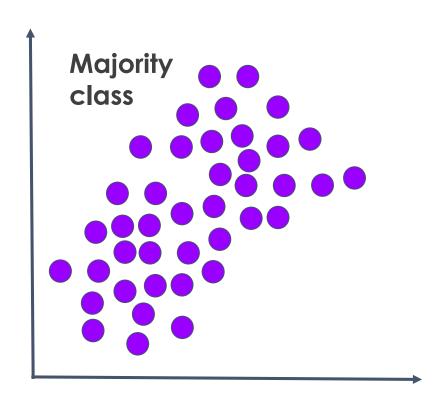


Use KNN to classify observations from majority class.





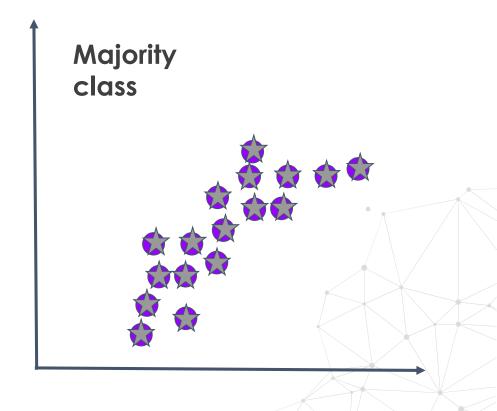






Retain observations that were wrongly classified.

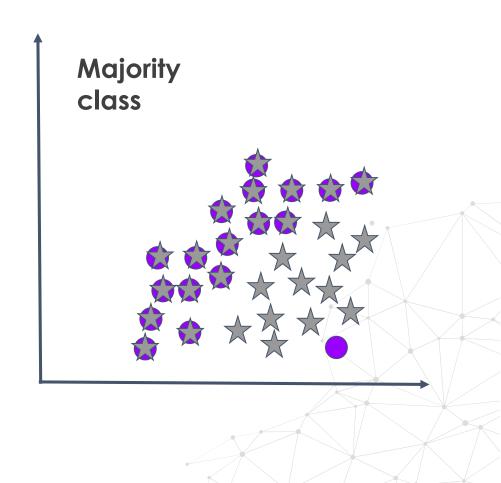








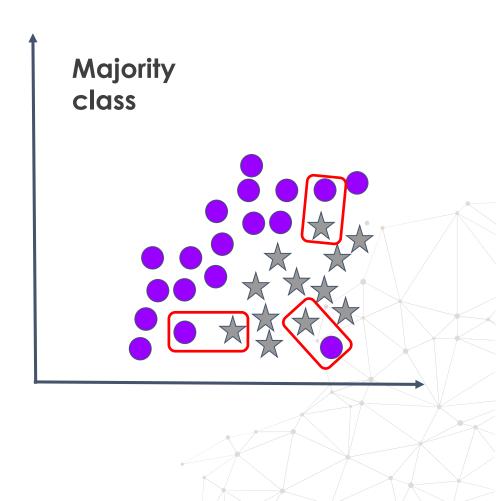
Resampled data consists of the minority group + those observations wrongly classified.







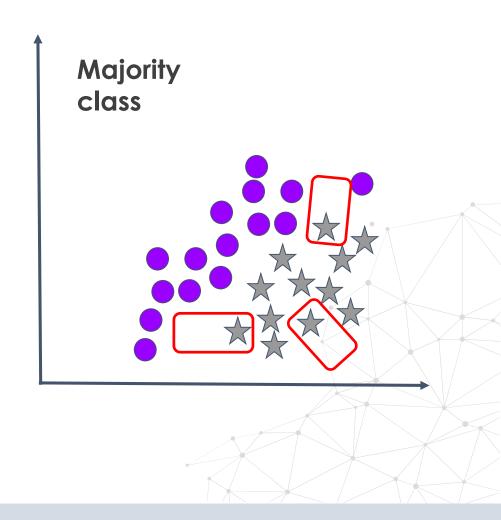
With the resampled data, find and remove the Tomek Links.







With the resampled data, find and remove the Tomek Links.



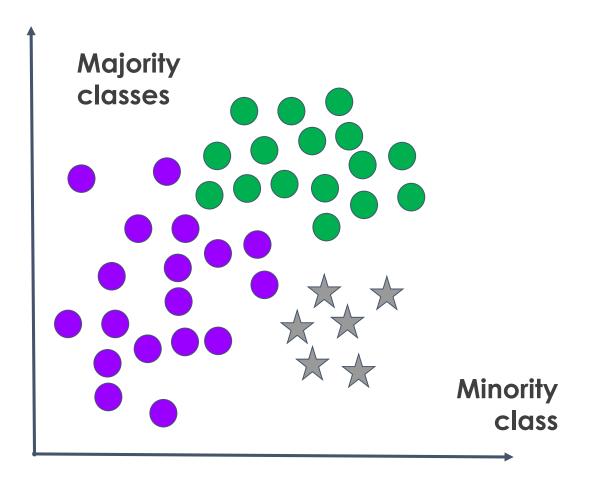


## Imbalanced-learn: OSS

```
# create data
X, y = make data(sep=2)
# set up OSS
oss = OneSidedSelection(
    sampling strategy='auto', # undersamples only the majority class
    random state=0, # for reproducibility
   n neighbors=1, # default
   n jobs=4) # I have 4 cores in my laptop
X resampled, y resampled = oss.fit resample(X, y)
```



## **Multi-class**

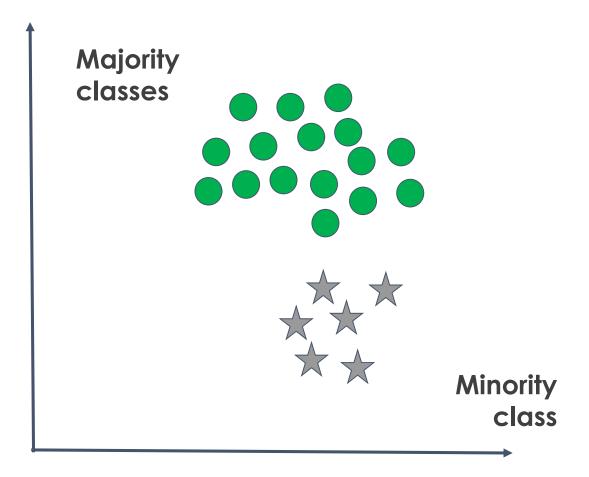


One vs One.





## **Multi-class**

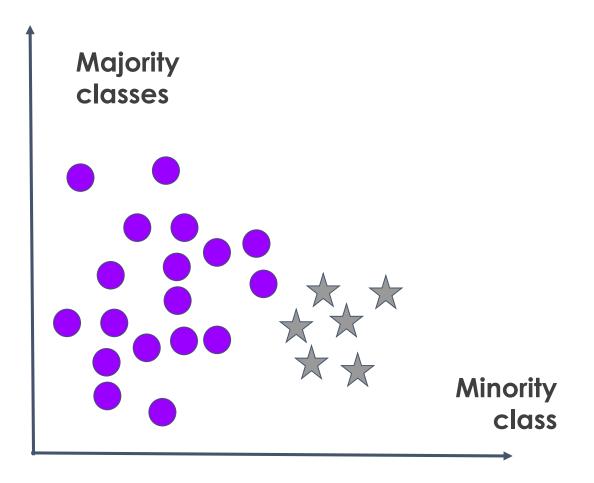


One vs One.

Run entire procedure over 1 majority class first



## **Multi-class**



One vs One.

Repeat the procedure for the other majority class.





# THANK YOU

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