

ASSIGNMENT 2

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GROUP 20

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datasets

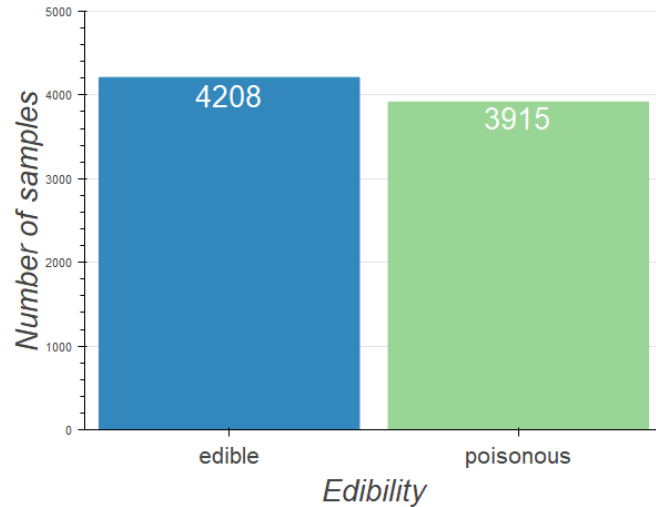
2

- CONGRESS
- WINE
- AMAZON
- MUSHROOMS

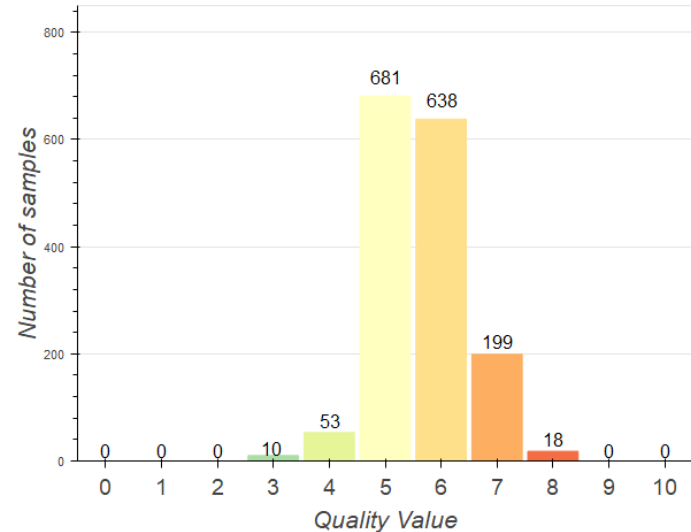
distribution of classes

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Distribution Mushrooms



Distribution Wine Quality



datasets overview

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Dataset	# samples	# attributes	Missing values?	# classes	classes eq. important?	Types of attributes
Amazon	750	10 000	no	50	yes	ordinal
Congress	217	18	yes (unknown)	2	yes	categorical (2 values)
Wine	1599	11	no	11	no	numeric
Mushrooms	8124	22	yes	2	no	categorical (2-10)

methods

- K-Nearest-Neighbors
- Support Vector Machines
- Random Forest

pre-processing

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scaling and selection

- Min-Max-Scaling
- Z-Score-Scaling
- Feature Selection
- Custom weighted scaling

encoding

- Ordinal Encoding
- One Hot Encoding
- TF-IDF

pre-processing

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Weighted scaling

If correlation < 0:

- $$\frac{\max(\text{feature}) - \text{feature}}{\max(\text{feature}) * |\text{correlation}|}$$

If correlation > 0:

- $$\frac{\text{feature}}{\max(\text{feature}) * |\text{correlation}|}$$

TF-IDF (text frequency - inverse document frequency)

- Upweighting rare words (important)
- Downweighting frequent words (syntactic)

$$\text{tfidf}(w, R) = \ln\left(\frac{\# \text{reviews}}{1 + \# \text{reviews containing } w}\right) * n(w, R)$$

results

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Best result	<u>kNN</u>	<u>Random Forest</u>	<u>SVM</u>
Wine		perfectly	perfectly
Congress			97% accuracy
Mushroom	perfectly	perfectly	perfectly
Amazon		67% accuracy	

conclusion

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different datasets

- kNN disappointing
 - high hopes kNN + TF-IDF
- SVM highly dependent on kernel
- Don't forget human intuition

Different methods

- SVM & RF outperformed
- kNN troubles with many classes
- only RF can handle heterogeneous data