Citation Request:

This dataset is public available for research. The details are described in [Cortez et al., 2009].

Please include this citation if you plan to use this database:

P. Cortez, A. Cerdeira, F. Almeida, T. Matos and J. Reis.

Modeling wine preferences by data mining from physicochemical properties.

In Decision Support Systems>, Elsevier, 47(4):547-553. ISSN: 0167-9236.

Available at: [@Elsevier] http://dx.doi.org/10.1016/j.dss.2009.05.016

[Pre-press (pdf)] http://www3.dsi.uminho.pt/pcortez/winequality09.pdf

[bib] http://www3.dsi.uminho.pt/pcortez/dss09.bib

1. Title: Wine Quality

2. Sources

Created by: Paulo Cortez (Univ. Minho), António Cerdeira, Fernando Almeida, Telmo Matos and José Reis (CVRVV) @ 2009

3. Past Usage:

P. Cortez, A. Cerdeira, F. Almeida, T. Matos and J. Reis.

Modeling wine preferences by data mining from physicochemical properties.

In Decision Support Systems>, Elsevier, 47(4):547-553. ISSN: 0167-9236.

In the above reference, two datasets were created, using red and white wine samples.

The inputs include objective tests (e.g. PH values) and the output is based on sensory data

(median of at least 3 evaluations made by wine experts). Each expert graded the wine quality

between 0 (very bad) and 10 (very excellent). Several data mining methods were applied to model

these datasets under a regression approach. The support vector machine model achieved the

best results. Several metrics were computed: MAD, confusion matrix for a fixed error tolerance (T),

etc. Also, we plot the relative importances of the input variables (as measured by a sensitivity

analysis procedure).

4. Relevant Information:

These datasets can be viewed as classification or regression tasks.

The classes are ordered and not balanced (e.g. there are munch more normal wines than

excellent or poor ones). Outlier detection algorithms could be used to detect the few excellent

or poor wines. Also, we are not sure if all input variables are relevant. So

it could be interesting to test feature selection methods.

5. Number of Instances: red wine - 1599; white wine - 4898.

6. Number of Attributes: 11 + output attribute

Note: several of the attributes may be correlated, thus it makes sense to apply some sort of

feature selection.

7. Attribute information:

For more information, read [Cortez et al., 2009].

Input variables (based on physicochemical tests):

1 - fixed acidity

2 - volatile acidity

3 - citric acid

4 - residual sugar

5 - chlorides

6 - free sulfur dioxide

7 - total sulfur dioxide

8 - density

9 - pH

10 - sulphates

11 - alcohol

Output variable (based on sensory data):

12 - quality (score between 0 and 10)

8. Missing Attribute Values: None