## Name of dataset: Breast Cancer Wisconsin (Diagnostic) Data Set

(Predict whether the cancer is benign or malignant)

## **About Dataset**

Features are computed from a digitized image of a fine needle aspirate (FNA) of a breast mass. They describe characteristics of the cell nuclei present in the image.

n the 3-dimensional space is that described in: [K. P. Bennett and O. L. Mangasarian: "Robust Linear Programming Discrimination of Two Linearly Inseparable Sets", Optimization Methods and Software 1, 1992, 23-34].

This database is also available through the UW CS ftp server:

ftp ftp.cs.wisc.edu

cd math-prog/cpo-dataset/machine-learn/WDBC/

Also can be found on UCI Machine Learning

Repository: <a href="https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+%28Diagnostic%29">https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+%28Diagnostic%29</a>
Attribute Information:

- 1) ID number
- 2) Diagnosis (M = malignant, B = benign)

3-32)

Ten real-valued features are computed for each cell nucleus:

- a) radius (mean of distances from center to points on the perimeter)
- b) texture (standard deviation of gray-scale values)
- c) perimeter
- d) area
- e) smoothness (local variation in radius lengths)
- f) compactness (perimeter^2 / area 1.0)
- g) concavity (severity of concave portions of the contour)
- h) concave points (number of concave portions of the contour)
- i) symmetry
- j) fractal dimension ("coastline approximation" 1)

The mean, standard error and "worst" or largest (mean of the three largest values) of these features were computed for each image, resulting in 30 features. For instance, field 3 is Mean Radius, field 13 is Radius SE, field 23 is Worst Radius.

All feature values are recoded with four significant digits.

Missing attribute values: none

Class distribution: 357 benign, 212 malignant

The number of samples used in training, validation, and testing depends on the split ratio specified in the **train\_test\_split** function. In the provided code, the test size is set to 0.30, which means 30% of the data will be used for testing, and the remaining 70% will be used for training.

Here's the breakdown of the samples:

- Training set: 70% of the total data
- Testing set: 30% of the total data

Given that the total number of samples (full dataset size) is 570, and the test size is set to 0.30 (30%), we can calculate the number of samples in each set:

- 1. Training set:  $570 \times 0.70 = 399570 \times 0.70 = 399$  samples
- 2. Testing set:  $570 \times 0.30 = 171 | 570 \times 0.30 = 171 |$  samples

So, in this scenario:

- The training set contains 399 samples.
- The testing set contains 171 samples.

## Implementation details:

Number of feature:30

## Names of feature:

- 1. radius\_mean
- 2. texture\_mean
- 3. perimeter\_mean
- 4. area\_mean
- 5. smoothness\_mean
- 6. compactness\_mean
- 7. concavity\_mean
- 8. concave points\_mean
- 9. symmetry\_mean
- 10. fractal\_dimension\_mean
- 11. radius se
- 12. texture\_se
- 13. perimeter\_se
- 14. area\_se
- 15. smoothness\_se
- 16. compactness\_se
- 17. concavity\_se
- 18. concave points\_se
- 19. symmetry\_se
- 20. fractal\_dimension\_se
- 21. radius\_worst
- 22. texture worst
- 23. perimeter\_worst
- 24. area worst
- 25. smoothness\_worst
- 26. compactness\_worst
- 27. concavity\_worst
- 28. concave points\_worst
- 29. symmetry\_worst
- 30. fractal\_dimension\_worst