

General information on dataset:

1: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: <https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+%28Diagnostic%29>

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 ($\text{perimeter}^2 / \text{area} - 1.0$)
- 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

2: The number of columns: 33, 30 feature, 1 classification

3: number of samples :570 sample

4: the number of samples used in training and testing.

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: $570 \times 0.70 = 399$ samples

2. Testing set: $570 \times 0.30 = 171$ 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

- **Batch Size:** Batch size is set to 1, as specified in `model.fit(X_train, y_train, batch_size=1, epochs=50)`.
- **Number of Epochs:** 50 epochs are used for training the model.

Additionally:

- **Loss Function:** Binary cross-entropy loss is used, specified as `loss="binary_crossentropy"`.
- **Metrics:** Model performance is evaluated using accuracy, specified as `metrics=["accuracy"]`.



