THEORY OF COMPUTATION

BITE306L

DIGITAL ASSIGNMENT - 1



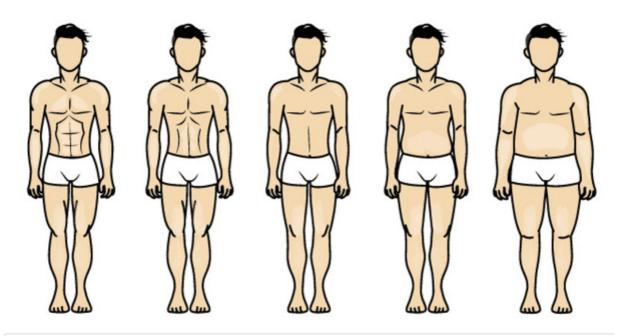
SUBMITTED BY,

KALIRAJ.A - 21BIT0687

GUIDED BY,

Dr. CHANDRASEGAR.T,
Department of Computer Applications,
School of Information Technology and Engineering,
VIT University, Vellore,
Tamil Nadu - 632014

DATASET LINK - https://www.kaggle.com/datasets/fedesoriano/body-fat-prediction-dataset



DATASET DESCRIPTION:

- ---> Lists estimates of the percentage of body fat determined by underwater weighing and various body circumference measurements for 200 men.
- ---> This data set can be used to illustrate multiple regression techniques. Accurate measurement of body fat is inconvenient/costly and it is desirable to have easy methods of estimating body fat that are not inconvenient/costly.
- ---> The variables listed below, from left to right, are:
 - 1)Density determined from underwater weighing
 - 2)Percent body fat from Siri's (1956) equation
 - 3)Age (years)
 - 4)Weight (lbs)
 - 5)Height (inches)
 - 6) Neck circumference (cm)
 - 7)Chest circumference (cm)
 - 8) Abdomen 2 circumference (cm)
 - 9) Hip circumference (cm)
 - 10)Thigh circumference (cm)

 - 11) Knee circumference (cm)
 - 12)Ankle circumference (cm)
 - 13)Biceps (extended) circumference (cm)
 - 14) Forearm circumference (cm)
 - 15)Wrist circumference (cm)
- ---> Number of columns : 15
- ---> Number of Rows : 200
- ---> Data Set Characteristics : Multivarite
- ---> Associated Tasks : Classification

---> Area : Medical - Health

References:

Bailey, Covert (1994). Smart Exercise: Burning Fat, Getting Fit, Houghton-Mifflin Co., Boston, pp. 179-186.

Behnke, A.R. and Wilmore, J.H. (1974). Evaluation and Regulation of Body Build and Composition, Prentice-Hall, Englewood Cliffs, N.J.

Siri, W.E. (1956), "Gross composition of the body", in Advances in Biological and Medical Physics, vol. IV, edited by J.H. Lawrence and C.A. Tobias, Academic Press, Inc., New York.

Katch, Frank and McArdle, William (1977). Nutrition, Weight Control, and Exercise, Houghton Mifflin Co., Boston.

Wilmore, Jack (1976). Athletic Training and Physical Fitness: Physiological Principles of the Conditioning Process, Allyn and Bacon, Inc., Boston.

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In [8]: import pandas as pd
        from sklearn.model selection import train test split
        from pandas.plotting import scatter_matrix
        import matplotlib.pyplot as plt
        from sklearn import model_selection
        from sklearn import metrics
        from sklearn.metrics import classification_report
        from sklearn.metrics import confusion_matrix
        from sklearn.metrics import accuracy score
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.tree import DecisionTreeClassifier
        import matplotlib.pyplot as plt
        import seaborn as sns
        print("KALIRAJ.A - 21BIT0687")
        data = pd.read_csv('bodyfat3.csv')
        # Split-out validation dataset
        array = data.values
        X = array[:,0:199]
        y = array[:,14]
        feature_cols = ['Density','BodyFat','Age','Weight','Height','Neck','Chest','Andomen','Hip
        # Split dataset into training set and test set
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3) # 70% training a
        #Create a Gaussian Classifier
        clf=RandomForestClassifier(n_estimators=100)
        #Train the model using the training sets y_pred=clf.predict(X_test)
        clf.fit(X_train,y_train)
        y_pred=clf.predict(X_test)
        # Model Accuracy, how often is the classifier correct?
        print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
        RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                    max_depth=None, max_features='auto', max_leaf_nodes=None,
                    min_impurity_decrease=0.0, min_impurity_split=None,
                    min_samples_leaf=1, min_samples_split=2,
                    min_weight_fraction_leaf=0.0, n_estimators=100, n_jobs=1,
                    oob_score=False, random_state=None, verbose=0,
                    warm_start=False)
        feature_imp = pd.Series(clf.feature_importances_,index=feature_cols).sort_values(ascendin
        feature imp
        %matplotlib inline
        # Creating a bar plot
        plt.figure()
        sns.barplot(x=feature_imp, y=feature_imp.index)
        # Add labels to your graph
        plt.xlabel('Feature Importance Score')
        plt.ylabel('Features')
        plt.title("Visualizing Important Features in dataset - Bodyfat")
```

plt.show()

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Accuracy: 0.93333333333333333

