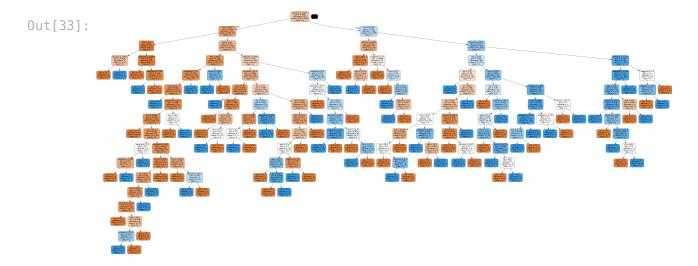
Praticle 5: Implementation of Decision Tree Classification Algorithms

```
In [30]: import pandas as pd
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.model selection import train test split
         from sklearn import metrics
         from sklearn.preprocessing import LabelEncoder
         # Load dataset
         col names = ['pregnant', 'glucose', 'bp', 'skin', 'insulin', 'bmi', 'pedi
         pima = pd.read_csv("diabetes.csv", header=0, names=col_names)
         # Preprocessing: Encode label column
         label encoder = LabelEncoder()
         pima['label'] = label encoder.fit transform(pima['label'])
         # Split dataset into features (X) and target (y)
         X = pima.drop('label', axis=1)
         y = pima['label']
In [31]: # 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,
         # Create Decision Tree classifier object
         clf = DecisionTreeClassifier()
         # Train Decision Tree classifier
         clf = clf.fit(X train, y train)
         # Predict the response for test dataset
         y pred = clf.predict(X test)
In [32]: # Evaluate the model
         print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
        Accuracy: 0.7056277056277056
In [33]: from sklearn.tree import export graphviz
         from six import StringIO
         from IPython.display import Image
         import pydotplus
         dot_data = StringIO()
         export graphviz(clf, out file=dot data,
                         filled=True, rounded=True,
                         special characters=True, feature names=col names[:-1], cl
         graph = pydotplus.graph from dot data(dot data.getvalue())
         graph.write png('diabetes.png')
         Image(graph.create png())
```



```
Optimizing Decision Tree Performance
In [34]: # Create Decision Tree classifer object
         clf = DecisionTreeClassifier(criterion="entropy", max_depth=3)
         # Train Decision Tree Classifer
         clf = clf.fit(X_train,y_train)
         #Predict the response for test dataset
         y_pred = clf.predict(X_test)
In [35]:
         # Model Accuracy, how often is the classifier correct?
         print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
        Accuracy: 0.7705627705627706
In [36]: from sklearn.tree import export_graphviz
         from six import StringIO
          from IPython.display import Image
         import pydotplus
         dot data = StringIO()
         export_graphviz(clf, out_file=dot_data,
                          filled=True, rounded=True,
                          special_characters=True, feature_names=col_names[:-1], cl
         graph = pydotplus.graph from dot data(dot data.getvalue())
         graph.write_png('diabetes.png')
         Image(graph.create png())
Out[36]:
                                                       bmi ≤ 28.15
ntropy = 0.977
amples = 195
lue = [80, 115]
```

samples = 18 value = [9, 9] class = 0

samples = 96 value = [41, 55]

Some Steps Were Followed to create and optimize the tree:

- The necessary libraries and modules are imported, including export_graphviz for visualizing the decision tree, StringIO for creating a file-like object in memory, Image for displaying the image in Jupyter Notebook, and pydotplus for handling the graph data.
- The dataset is loaded from a CSV file into a pandas DataFrame, with column names specified.
- The dataset is split into training and test sets using the train_test_split function from scikit-learn. 70% of the data is used for training, and 30% is reserved for testing.
- A decision tree classifier object is created using the default settings.
- The decision tree classifier is trained on the training data using the fit method.
- The classifier is used to predict the labels for the test dataset using the predict method.
- The accuracy of the model is calculated by comparing the predicted labels with the actual labels from the test dataset.
- The decision tree is visualized using the export_graphviz function to generate the
 dot data, which is then converted into a PNG image using pydotplus. The image is
 saved as "diabetes.png" and displayed using Image.
- The decision tree classifier is redefined with some additional settings, such as using entropy as the criterion and setting a maximum depth of 3.
- Steps 5-8 are repeated with the updated decision tree classifier to visualize the new tree and calculate its accuracy.