```
import pandas as pd
    import matplotlib.pyplot as plt
    from sklearn.model_selection import train_test_split
    from \ sklearn.tree \ import \ Decision Tree Classifier
    from sklearn.ensemble import RandomForestClassifier
    \verb|wrl = "https://raw.githubusercontent.com/jbrownlee/Datasets/master/pima-indians-diabetes.data.csv"|
    columns = ["Pregnncies", "Glucose", "BloodPressure", "SkinThickness", "Insulin", "BMI", "DiabetesPedigreeFunction", "Age", "Outcom
    df = pd.read_csv(url, names = columns)
    df
          Pregnncies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
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    768 rows × 9 columns
Next steps: ( Generate code with df ) ( View recommended plots )
                                                                     New interactive sheet
    x = df.drop("Outcome",axis =1)
    y = df["Outcome"]
    x_train, x_test, y_train, y_test = train_test_split(x,y,test_size =0.2, random_state = 42)
    dt_clf = DecisionTreeClassifier(criterion = 'entropy', max_depth = 3, random_state = 42)
    dt_clf.fit(x_train, y_train)
                              DecisionTreeClassifier
    DecisionTreeClassifier(criterion='entropy', max_depth=3, random_state=42)
    rf_clf = RandomForestClassifier(n_estimators = 100, random_state = 42)
    rf_clf.fit(x_train, y_train)
            RandomForestClassifier
    RandomForestClassifier(random state=42)
    \label{eq:print}  \text{print}(\texttt{"Decision Tree - Training Accuracy: ", dt_clf.score}(x\_train, y\_train)) \\
    print("Decision Tree - Testing Accuracy: ", dt_clf.score(x_test, y_test))
    Decision Tree - Training Accuracy: 0.7752442996742671
    Decision Tree - Testing Accuracy: 0.7662337662337663
    print("Random Forest - Training Accuracy: ", rf_clf.score (rain, y_train))
print("Decision Tree - Testing Accuracy: ", rf_clf.score (st, y_test))
```

```
Random Forest - Training Accuracy: 1.0
Decision Tree - Testing Accuracy: 0.7207792207792207
```

```
importances = rf_clf.feature_importances_
feat_importance = pd.Series(importances, index =x.columns).sort_values(ascending = False)
```

```
plt.figure(figsize=(8,5))
feat_importance.plot(kind='bar', color = "skyblue")
plt.ylabel(" Feature Importance - Random Forest")
plt.ylabel("Importance Score")
plt.show()
```

