

IMPORTING THE REQUIRED PACKAGES

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, precision_score, recall_score
```

Load the dataset

```
df = pd.read_csv('/content/diabetes.csv') # Update path if needed
print("Dataset loaded successfully!")
print(df.head())
```

```
↻ Dataset loaded successfully!
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI \
0	6	148	72	35	0	33.6
1	1	85	66	29	0	26.6
2	8	183	64	0	0	23.3
3	1	89	66	23	94	28.1
4	0	137	40	35	168	43.1

	DiabetesPedigreeFunction	Age	Outcome
0	0.627	50	1
1	0.351	31	0
2	0.672	32	1
3	0.167	21	0
4	2.288	33	1

Prepare data (features and target)

```
X = df.drop('Outcome', axis=1)
y = df['Outcome']
```

Split data into training and testing

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
print("Train-test split completed.")
```

```
↻ Train-test split completed.
```

Standardize the data

```
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

Train and evaluate multiple models

```
# Initialize models
models = {
    "Logistic Regression": LogisticRegression(),
    "Random Forest": RandomForestClassifier(),
    "Support Vector Machine": SVC(probability=True)
}
```

```
# Initialize empty lists for metrics
model_names = []
accuracies = []
precisions = []
recalls = []
```

```
# Train and evaluate models
for name, model in models.items():
```

```

model.fit(X_train, y_train)
predictions = model.predict(X_test)

acc = accuracy_score(y_test, predictions)
prec = precision_score(y_test, predictions)
rec = recall_score(y_test, predictions)

model_names.append(name)
accuracies.append(acc)
precisions.append(prec)
recalls.append(rec)

# Create a DataFrame for results
results = pd.DataFrame({
    "Model": model_names,
    "Accuracy": accuracies,
    "Precision": precisions,
    "Recall": recalls
})

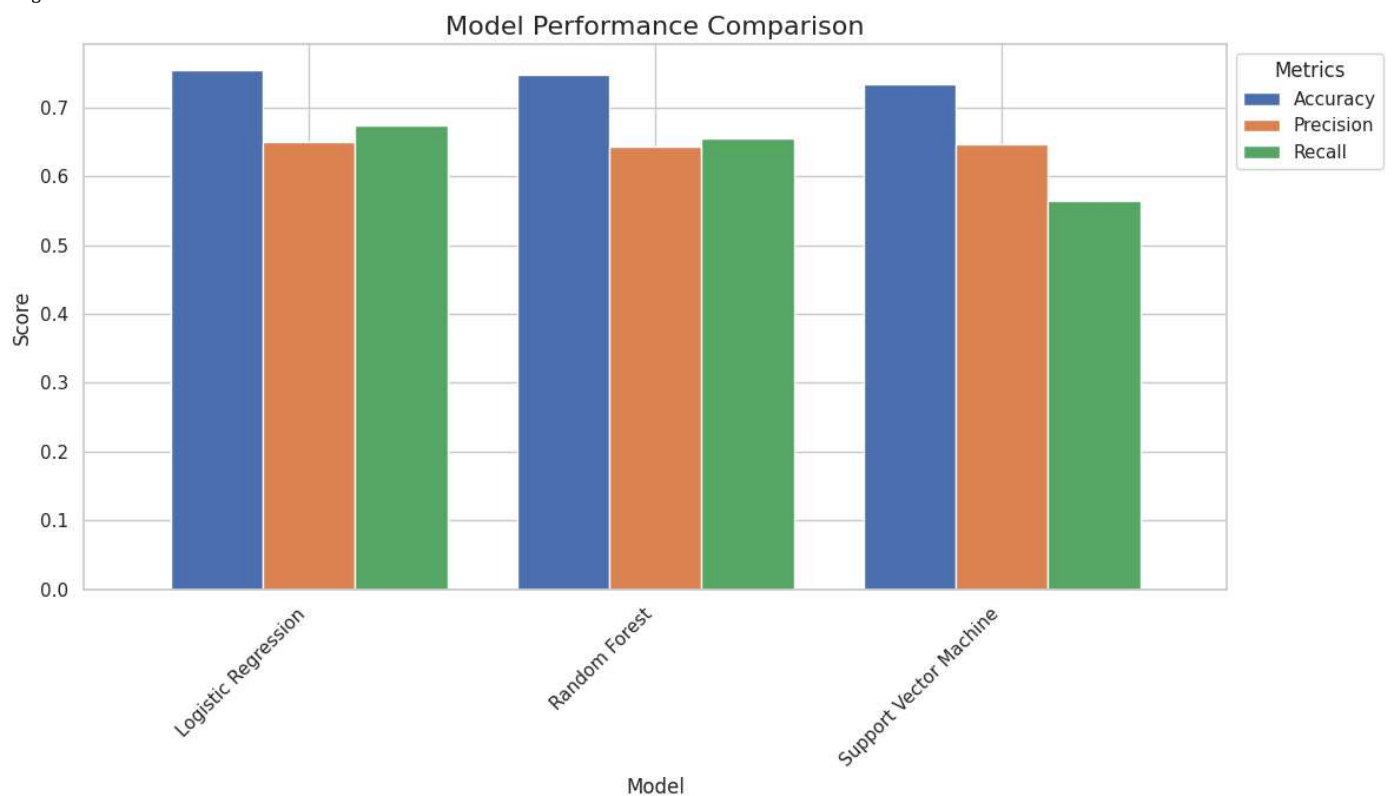
# Plot the results using Seaborn
plt.figure(figsize=(12, 7))
sns.set(style="whitegrid")

# Plot the bar chart for model comparison
results.set_index('Model').plot(kind='bar', figsize=(12, 7), width=0.8)
plt.title('Model Performance Comparison', fontsize=16)
plt.ylabel('Score', fontsize=12)
plt.xlabel('Model', fontsize=12)
plt.xticks(rotation=45, ha="right")
plt.legend(title='Metrics', loc='upper left', bbox_to_anchor=(1, 1))
plt.tight_layout()

# Show the plot
plt.show()

```

↗ <Figure size 1200x700 with 0 Axes>



Showing best Model output

```

import matplotlib.pyplot as plt
import numpy as np

```

```

# Assuming you have the data for accuracies, precisions, recalls, and model_names
accuracies = [0.75, 0.80, 0.76] # Example accuracy values for models
precisions = [0.74, 0.79, 0.75] # Example precision values for models
recalls = [0.76, 0.78, 0.77] # Example recall values for models
model_names = ['Logistic Regression', 'Random Forest', 'SVM'] # Example model names

# Finding the best model based on Accuracy
best_model = model_names[np.argmax(accuracies)]
print(f"Best Model: {best_model} with Accuracy: {max(accuracies):.4f}")

# Set positions for each bar (since we're comparing multiple models and metrics)
x = np.arange(len(model_names)) # The label locations
width = 0.2 # Width of the bars

# Create a grouped bar chart for the models' performance metrics (Accuracy, Precision, Recall)
fig, ax = plt.subplots(figsize=(10, 6))

# Plot bars for each metric
bars_accuracy = ax.bar(x - width, accuracies, width, label='Accuracy', color='lightblue')
bars_precision = ax.bar(x, precisions, width, label='Precision', color='lightgreen')
bars_recall = ax.bar(x + width, recalls, width, label='Recall', color='lightcoral')

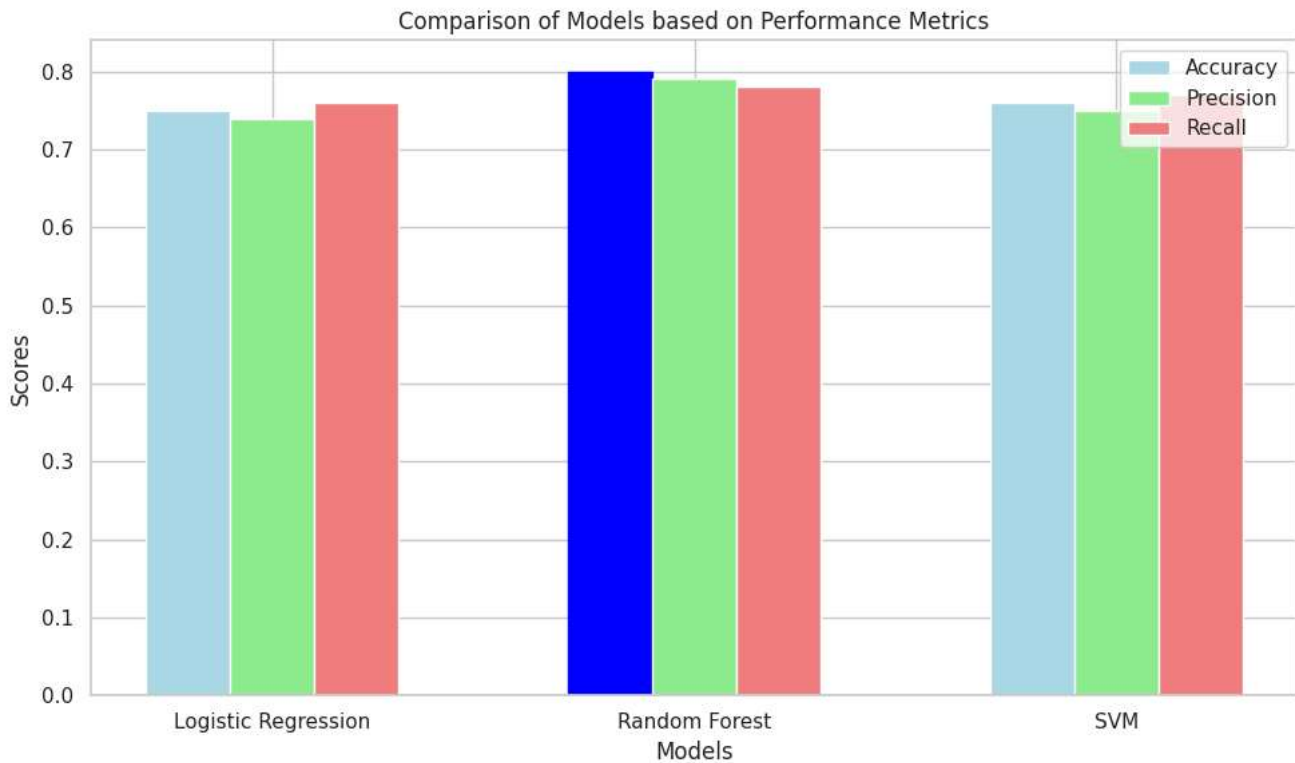
# Add labels, title, and customize the plot
ax.set_xlabel('Models')
ax.set_ylabel('Scores')
ax.set_title('Comparison of Models based on Performance Metrics')
ax.set_xticks(x)
ax.set_xticklabels(model_names)
ax.legend()

# Highlight the best model based on accuracy
best_index = np.argmax(accuracies)
bars_accuracy[best_index].set_color('blue') # Highlight the best model's accuracy bar in blue

plt.tight_layout()
plt.show()

```

➡ Best Model: Random Forest with Accuracy: 0.8000



precautions !!

```

def provide_suggestions_for_diabetic():
    print("\nPatient is diabetic. Here are some suggestions:")
    print("    Follow a diabetic friendly diet (low sugar and carbs) ")

```

```
print( - Follow a diabetic-friendly diet (low sugar and carbs). )
print("- Engage in regular physical activity (e.g., walking, swimming).")
print("- Monitor blood sugar levels regularly.")
print("- Take prescribed medications as directed by a healthcare professional.")
print("- Get regular checkups and screenings.")
```

```
provide_suggestions_for_diabetic()
```



Patient is diabetic. Here are some suggestions:

- Follow a diabetic-friendly diet (low sugar and carbs).
- Engage in regular physical activity (e.g., walking, swimming).
- Monitor blood sugar levels regularly.
- Take prescribed medications as directed by a healthcare professional.
- Get regular checkups and screenings.