

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, classification_report
```

Load Dataset

```
In [2]: data = load_breast_cancer()
df = pd.DataFrame(data.data, columns=data.feature_names)
df['target'] = data.target
```

Data Preprocessing

```
In [3]: scaler = StandardScaler()
X = scaler.fit_transform(df.drop(columns=['target']))
y = df['target']
```

Split Data

```
In [4]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Initialize Models

```
In [5]: models = {
    "Logistic Regression": LogisticRegression(),
    "Decision Tree": DecisionTreeClassifier(),
    "Random Forest": RandomForestClassifier(),
    "SVM": SVC(),
    "k-NN": KNeighborsClassifier()
}
```

Train and Evaluate Models

In [6]:

```
results = {}  
for name, model in models.items():  
    model.fit(X_train, y_train)  
    y_pred = model.predict(X_test)  
    accuracy = accuracy_score(y_test, y_pred)  
    results[name] = accuracy  
    print(f"{name} Accuracy: {accuracy:.4f}")  
    print(classification_report(y_test, y_pred))  
    print("-"*50)
```

```
Logistic Regression Accuracy: 0.9737
      precision    recall  f1-score   support

     0       0.98      0.95      0.96         43
     1       0.97      0.99      0.98         71

 accuracy          0.97         114
 macro avg          0.97         114
weighted avg          0.97         114
```

```
-----
Decision Tree Accuracy: 0.9474
      precision    recall  f1-score   support

     0       0.93      0.93      0.93         43
     1       0.96      0.96      0.96         71

 accuracy          0.95         114
 macro avg          0.94         114
weighted avg          0.95         114
```

```
-----
Random Forest Accuracy: 0.9649
      precision    recall  f1-score   support

     0       0.98      0.93      0.95         43
     1       0.96      0.99      0.97         71

 accuracy          0.96         114
 macro avg          0.97         114
weighted avg          0.97         114
```

```
-----
SVM Accuracy: 0.9737
      precision    recall  f1-score   support

     0       0.98      0.95      0.96         43
     1       0.97      0.99      0.98         71

 accuracy          0.97         114
 macro avg          0.97         114
weighted avg          0.97         114
```

```
-----
k-NN Accuracy: 0.9474
      precision    recall  f1-score   support

     0       0.93      0.93      0.93         43
     1       0.96      0.96      0.96         71

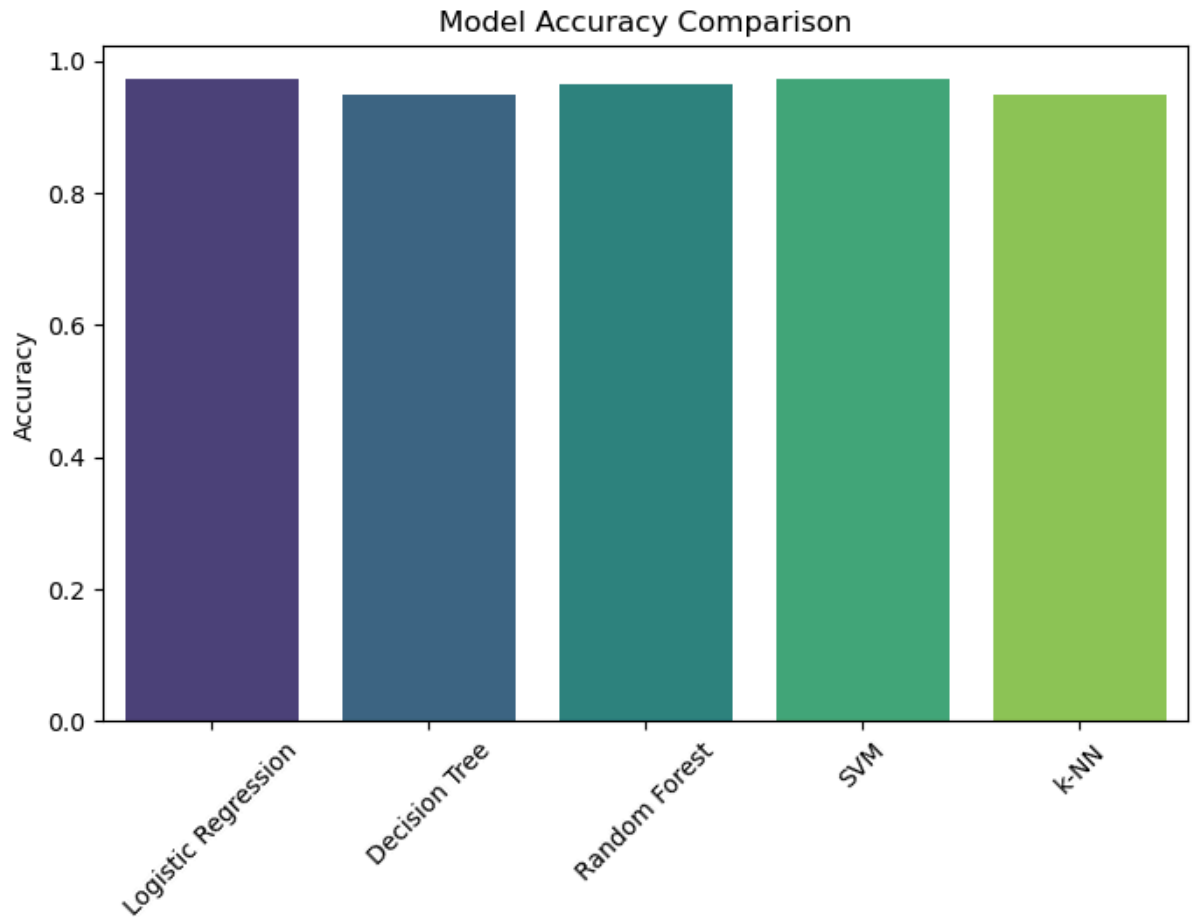
 accuracy          0.95         114
 macro avg          0.94         114
weighted avg          0.95         114

-----
```

Compare Model Performance

In [7]:

```
plt.figure(figsize=(8, 5))
sns.barplot(x=list(results.keys()), y=list(results.values()), palette='viridis')
plt.title("Model Accuracy Comparison")
plt.xticks(rotation=45)
plt.ylabel("Accuracy")
plt.show()
```



Identify Best and Worst Performing Models

```
In [8]: best_model = max(results, key=results.get)
worst_model = min(results, key=results.get)
print(f"Best Performing Model: {best_model} with {results[best_model]:.4f} accuracy")
print(f"Worst Performing Model: {worst_model} with {results[worst_model]:.4f} accuracy")
```

Best Performing Model: Logistic Regression with 0.9737 accuracy
Worst Performing Model: Decision Tree with 0.9474 accuracy

In []: