

# classifying-irises

February 13, 2025

```
[1]: from sklearn.datasets import load_iris
```

```
[2]: X, y = load_iris(as_frame=True, return_X_y=True)
```

Scale the features as we did with the hierarchical clustering demo

```
[3]: from sklearn.preprocessing import StandardScaler
```

```
[4]: scaler = StandardScaler()  
X_scaled = scaler.fit_transform(X)  
X_scaled
```

```
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```

Build our decision tree model for the purposes of extracting the feature importances which we can use to update the feature weighting in the hierarchical clustering demo

```
[5]: from sklearn.model_selection import train_test_split
```

```
[6]: X_train, X_test, y_train, y_test = train_test_split(
      X_scaled,
      y,
```

```
test_size=0.3,  
random_state=42)
```

```
[7]: from sklearn.tree import DecisionTreeClassifier  
from sklearn.ensemble import RandomForestClassifier
```

Tree-based classifiers like `DecisionTreeClassifier` and `RandomForestClassifier` give us weights indicating how important each feature is using `feature_importances_`

```
[8]: #clf = DecisionTreeClassifier()  
clf = RandomForestClassifier()  
  
clf = clf.fit(X_train, y_train)
```

```
[9]: y_pred = clf.predict(X_test)  
y_pred
```

```
[9]: array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,  
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          0])
```

```
[10]: from sklearn.metrics import accuracy_score
```

Test that our model is actually working

```
[11]: accuracy_score(y_test, y_pred)
```

```
[11]: 1.0
```

Get our feature importances and graph them for comparison

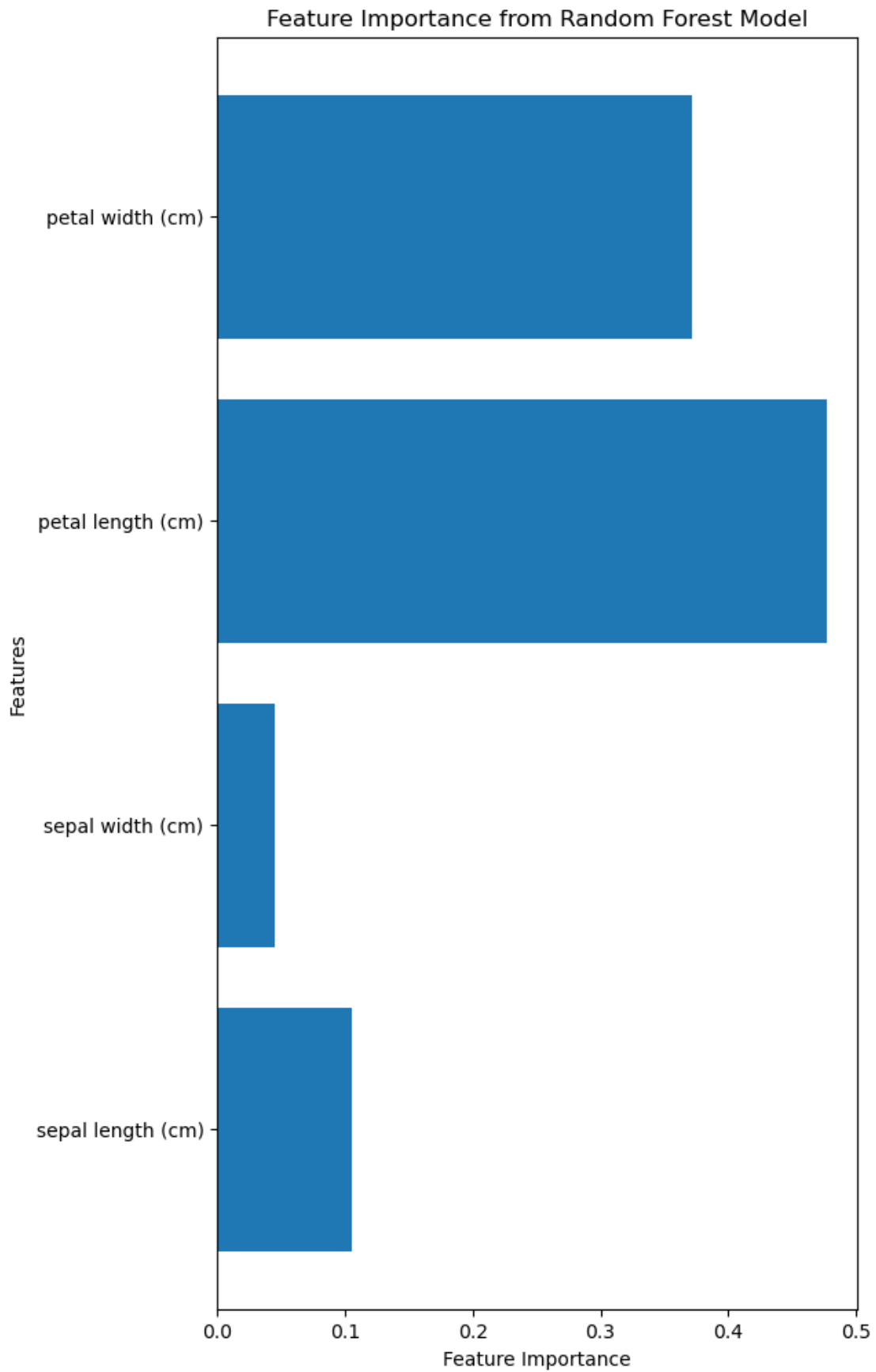
```
[12]: clf.feature_importances_
```

```
[12]: array([0.10587484, 0.04496103, 0.47740956, 0.37175456])
```

```
[13]: import matplotlib.pyplot as plt
```

```
[14]: plt.figure(figsize=(6, 12))  
plt.barh(X.columns, clf.feature_importances_)  
plt.xlabel('Feature Importance')  
plt.ylabel('Features')  
plt.title('Feature Importance from Random Forest Model')  
plt.show()
```





We can scale these so that the maximum value is 1, making it easier to compare their relative importance

```
[15]: clf.feature_importances_ / max(clf.feature_importances_)

[15]: array([0.22176942, 0.09417707, 1.          , 0.77869107])
```