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# Lab: Random Forest Algorithm
# 0. Installation and Imports

# Uncomment and run if not installed:
!pip install ucimlrepo

from ucimlrepo import fetch_ucirepo
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
import matplotlib.pyplot as plt
```

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Collecting ucimlrepo
  Downloading ucimlrepo-0.0.7-py3-none-any.whl.metadata (5.5 kB)
Requirement already satisfied: pandas>=1.0.0 in /usr/local/lib/python3.12/...
Requirement already satisfied: certifi>=2020.12.5 in /usr/local/lib/python3.12/...
Requirement already satisfied: numpy>=1.26.0 in /usr/local/lib/python3.12/...
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.12/...
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-...
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/...
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-...
Downloading ucimlrepo-0.0.7-py3-none-any.whl (8.0 kB)
Installing collected packages: ucimlrepo
Successfully installed ucimlrepo-0.0.7
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# Lab: Random Forest Algorithm
# 1. Load Dataset and Inspect

dermatology = fetch_ucirepo(id=33)

X = dermatology.data.features
y = dermatology.data.targets

print("Metadata:\n", dermatology.metadata)
print("\nVariables:\n", dermatology.variables)
print("Sample features:\n", X.head())
print("Sample targets:\n", y.head())
```

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disappearance of the granular layer \
0      0
1      0
2      0
3      3
4      2

vacuolisation and damage of the basal layer  spongiosis \
0      0      3
1      0      0
2      2      3
3      0      0
4      3      2

saw-tooth appearance of retes  follicular horn plug \
0      0      0
1      0      0
2      2      0
3      0      0
4      3      0

perifollicular parakeratosis  inflammatory monoluclear infiltrate \
0      0      1
1      0      1
2      0      2
3      0      3
4      0      2

band-like infiltrate  age
0      0  55.0
1      0   8.0
2      3  26.0
3      0  40.0
4      3  45.0

[5 rows x 34 columns]
Sample targets:
class
0      2
1      1
2      3
3      1
4      3

```

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# Lab: Random Forest Algorithm
# 2. Train-Test Split

X_train, X_test, y_train, y_test = train_test_split(
    X, y.values.ravel(), test_size=0.2, random_state=42
)

print("Training set size:", X_train.shape)
print("Testing set size:", X_test.shape)

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Training set size: (292, 34)
Testing set size: (74, 34)

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# Lab: Random Forest Algorithm
# 3. Random Forest Training

rf = RandomForestClassifier(n_estimators=100, random_state=42)
rf.fit(X_train, y_train)
y_pred = rf.predict(X_test)
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# Lab: Random Forest Algorithm
# 4. Evaluation and Output

print("Accuracy Score:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test, y_pred))
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

Accuracy Score: 0.9864864864864865

Classification Report:

	precision	recall	f1-score	support
1	1.00	1.00	1.00	31
2	0.90	1.00	0.95	9
3	1.00	1.00	1.00	13
4	1.00	0.88	0.93	8
5	1.00	1.00	1.00	10
6	1.00	1.00	1.00	3
accuracy			0.99	74
macro avg	0.98	0.98	0.98	74
weighted avg	0.99	0.99	0.99	74

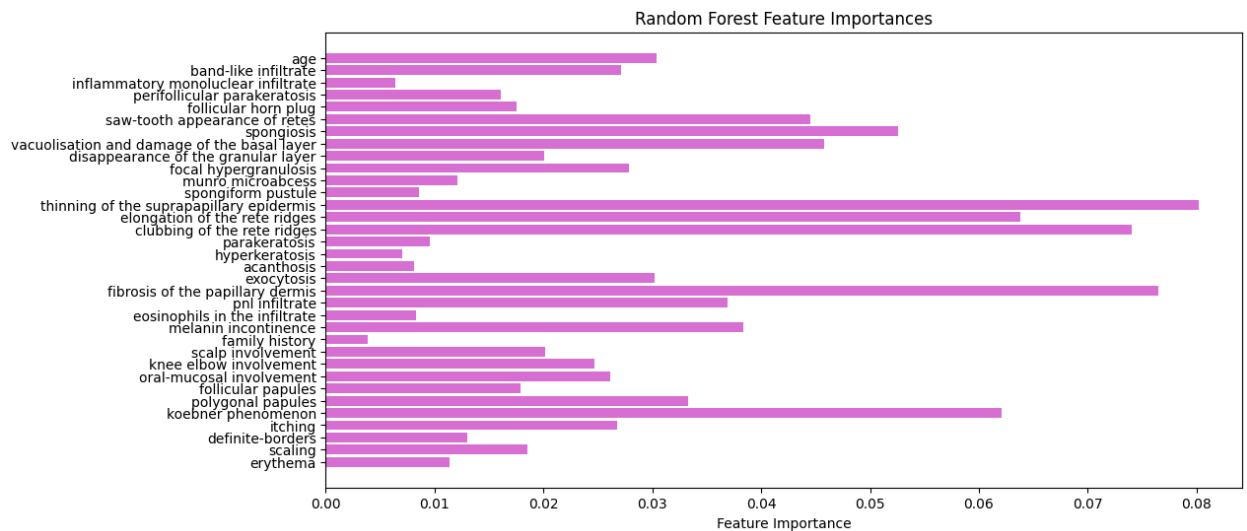
Confusion Matrix:

```
[[31  0  0  0  0  0]
 [ 0  9  0  0  0  0]
 [ 0  0 13  0  0  0]
 [ 0  1  0  7  0  0]
 [ 0  0  0  0 10  0]
 [ 0  0  0  0  0  3]]
```

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# Lab: Random Forest Algorithm
# 5. Feature Importance Visualization

importances = rf.feature_importances_
features = X.columns

plt.figure(figsize=(12, 6))
plt.barh(features, importances, color='orchid')
plt.xlabel('Feature Importance')
plt.title('Random Forest Feature Importances')
plt.show()
```



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