

2/03/25

Lab-2

Qn: Use an appropriate dataset for building the decision tree (ID3) & apply this knowledge to classify a new sample.

```
→ import pandas as pd
import numpy as np
from collections import Counter
import math
```

```
def entropy(y):
```

```
    counts = Counter(y)
```

```
    prob = [count/len(y) for count in counts.values()]
```

```
    return -sum(p * math.log2(p) for p in prob)
```

```
def info_gain(data, feature, target):
```

```
    total_entropy = entropy(data[target])
```

```
    values = data[feature].unique()
```

```
    weighted_entropy = sum((len(data[feature] == v)) /
                             len(data)) * entropy(data[data[feature] == v][target])
    for v in values)
```

```
    return total_entropy - weighted_entropy
```

```
def id3(data, features, target):
```

```
    if len(set(data[target])) == 1:
```

```
        return data[target].iloc[0]
```

```
    if len(features) == 0:
```

```
        return data[target].mode()[0]
```

```
    gains = {feature: info_gain(data, feature, target)
              for feature in features}
```

```
    best_feature = max(gains, key=gains.get)
```



```
tree = {best_feature: 1}
return tree
```

```
def print_tree(tree, indent = " "):
    if not isinstance(tree, dict):
        print(indent + "→" + str(tree))
        return
    for key, value in tree.items():
        print(indent + str(key))
        for sub_key, sub_tree in value.items():
            print(indent + "L" + str(sub_key))
```

```
path = '/content/Jennis.csv'
data = pd.read_csv(path)
features = list(data.columns[:-1])
target = 'play'
decision_tree = id3(data, features, target)
print_tree(decision_tree)
```

→ Output :-

outlook

├ Sunny

├ ── humidity

├ ── high: no

├ ── normal: yes

├ ── overcast: yes

├ rainy

├ ── windy

├ ── False: yes

├ ── True: no

Aug 13