

MACHINE LEARNING Laboratory

5th Semester, Academic Year 2025

Date:19-08-2025

| | | |
|--------------------|-------------------|--------------|
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|--------------------|-------------------|--------------|

Week# ____3____

MUSHROOMS.CSV

```
PS C:\Users\Dhruv Thakur\Downloads\all> python code\pytorch_implementation\test.py --ID EC_C_PES2UG23CS175_LAB3 --data mushrooms.csv --framework pytorch --print-tree --print-construct
Running tests with PYTORCH framework
=====
target column: 'class' (last column)
Original dataset info:
Shape: (8124, 23)
Columns: ['cap-shape', 'cap-surface', 'cap-color', 'bruises', 'odor', 'gill-attachment', 'gill-spacing', 'gill-size', 'gill-color', 'stalk-shape', 'stalk-root', 'stalk-surface-above-
ring', 'stalk-surface-below-ring', 'stalk-color-above-ring', 'stalk-color-below-ring', 'veil-type', 'veil-color', 'ring-number', 'ring-type', 'spore-print-color', 'population', 'hab
itat', 'class']

First few rows:

cap-shape: ['x' 'b' 's' 'f' 'k'] -> [5 0 4 2 3]

cap-surface: ['s' 'y' 'f' 'g'] -> [2 3 0 1]

cap-color: ['n' 'y' 'w' 'g' 'e'] -> [4 9 8 3 2]

class: ['p' 'e'] -> [1 0]

Processed dataset shape: torch.Size([8124, 23])
Number of features: 22
Features: ['cap-shape', 'cap-surface', 'cap-color', 'bruises', 'odor', 'gill-attachment', 'gill-spacing', 'gill-size', 'gill-color', 'stalk-shape', 'stalk-root', 'stalk-surface-above-
ring', 'stalk-surface-below-ring', 'stalk-color-above-ring', 'stalk-color-below-ring', 'veil-type', 'veil-color', 'ring-number', 'ring-type', 'spore-print-color', 'population', 'hab
itat']
Target: class
Framework: PYTORCH
Data type: <class 'torch.Tensor'>

=====
DECISION TREE CONSTRUCTION DEMO
=====
Total samples: 8124
Training samples: 6499
Testing samples: 1625
```

```
EXPLORER
ALL
code
  pytorch_implementation
  _pycache_
  EC_C_PES2UG23CS175_LAB3.py
  student_lab.py
  test.py
  sklearn_implementation
code.zip
mushrooms.csv
Nursery.csv
student_version.pdf
tictactoe.csv

PROBLEMS
OUTPUT
DEBUG CONSOLE
TERMINAL
PORTS

Constructing decision tree using training data...
Level 0: Node Info - Entropy = 0.9983
Level 0: Node Info - Selected Attribute: odor (gain: 0.9083)
Level 0: Node Info - Branch odor = 0
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 0
Level 0: Node Info - Branch odor = 1
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 1
Level 0: Node Info - Branch odor = 2
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 1
Level 0: Node Info - Branch odor = 3
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 0
Level 0: Node Info - Branch odor = 4
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 1
Level 0: Node Info - Branch odor = 5
Level 1: Node Info - Entropy = 0.2059
Level 1: Node Info - Selected Attribute: spore-print-color (gain: 0.1469)
Level 1: Node Info - Branch spore-print-color = 0
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 0
Level 1: Node Info - Branch spore-print-color = 1
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 0
Level 1: Node Info - Branch spore-print-color = 2
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 0
Level 1: Node Info - Branch spore-print-color = 3
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 0
Level 1: Node Info - Branch spore-print-color = 4
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 0
Level 1: Node Info - Branch spore-print-color = 5
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 1
Level 1: Node Info - Branch spore-print-color = 7
Level 2: Node Info - Entropy = 0.3333
Level 2: Node Info - Selected Attribute: habitat (gain: 0.2217)
Level 2: Node Info - Branch habitat = 0
Level 3: Node Info - Entropy = 0.7642
Level 3: Node Info - Selected Attribute: gill-size (gain: 0.7642)
Level 3: Node Info - Branch gill-size = 0
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 0
Level 3: Node Info - Branch gill-size = 1
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 1
Level 2: Node Info - Branch habitat = 1
Level 3: Node Info - Entropy = -0.0000
Level 3: Node Info - Hypothesis: Class 0
Level 2: Node Info - Branch habitat = 2
Level 3: Node Info - Entropy = -0.0000
Level 3: Node Info - Hypothesis: Class 1
```

```
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ALL
code
  pytorch_implementation
  _pycache_
  EC_C_PES2UG23CS175_LAB3.py
  student_lab.py
  test.py
  sklearn_implementation
code.zip
mushrooms.csv
Nursery.csv
student_version.pdf
tictactoe.csv

PROBLEMS
OUTPUT
DEBUG CONSOLE
TERMINAL
PORTS

Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 1
Level 1: Node Info - Branch spore-print-color = 7
Level 2: Node Info - Entropy = 0.3333
Level 2: Node Info - Selected Attribute: habitat (gain: 0.2217)
Level 2: Node Info - Branch habitat = 0
Level 3: Node Info - Entropy = 0.7642
Level 3: Node Info - Selected Attribute: gill-size (gain: 0.7642)
Level 3: Node Info - Branch gill-size = 0
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 0
Level 3: Node Info - Branch gill-size = 1
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 1
Level 2: Node Info - Branch habitat = 1
Level 3: Node Info - Entropy = -0.0000
Level 3: Node Info - Hypothesis: Class 0
Level 2: Node Info - Branch habitat = 2
Level 3: Node Info - Entropy = 0.7300
Level 3: Node Info - Selected Attribute: cap-color (gain: 0.7300)
Level 3: Node Info - Branch cap-color = 1
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 0
Level 3: Node Info - Branch cap-color = 4
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 0
Level 3: Node Info - Branch cap-color = 8
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 1
Level 3: Node Info - Branch cap-color = 9
Level 4: Node Info - Entropy = -0.0000
Level 4: Node Info - Hypothesis: Class 1
Level 2: Node Info - Branch habitat = 4
Level 3: Node Info - Entropy = -0.0000
Level 3: Node Info - Hypothesis: Class 0
Level 2: Node Info - Branch habitat = 6
Level 3: Node Info - Entropy = -0.0000
Level 3: Node Info - Hypothesis: Class 0
Level 1: Node Info - Branch spore-print-color = 8
Level 2: Node Info - Entropy = -0.0000
Level 2: Node Info - Hypothesis: Class 0
Level 0: Node Info - Branch odor = 6
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 1
Level 0: Node Info - Branch odor = 7
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 1
Level 0: Node Info - Branch odor = 8
Level 1: Node Info - Entropy = -0.0000
Level 1: Node Info - Hypothesis: Class 1

Decision tree construction completed using PYTORCH!

DECISION TREE STRUCTURE
Root [odor] (gain: 0.9083)
```

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EXPLORER

- ALL
- code
 - pytorch_implementation
 - __pycache__
 - EC_C_PES2UG23CS175_LAB3.py
 - student_lab.py
 - test.py
- sklearn_implementation
- code.zip
- mushrooms.csv
- Nursery.csv
- student_version.pdf
- tictactoe.csv

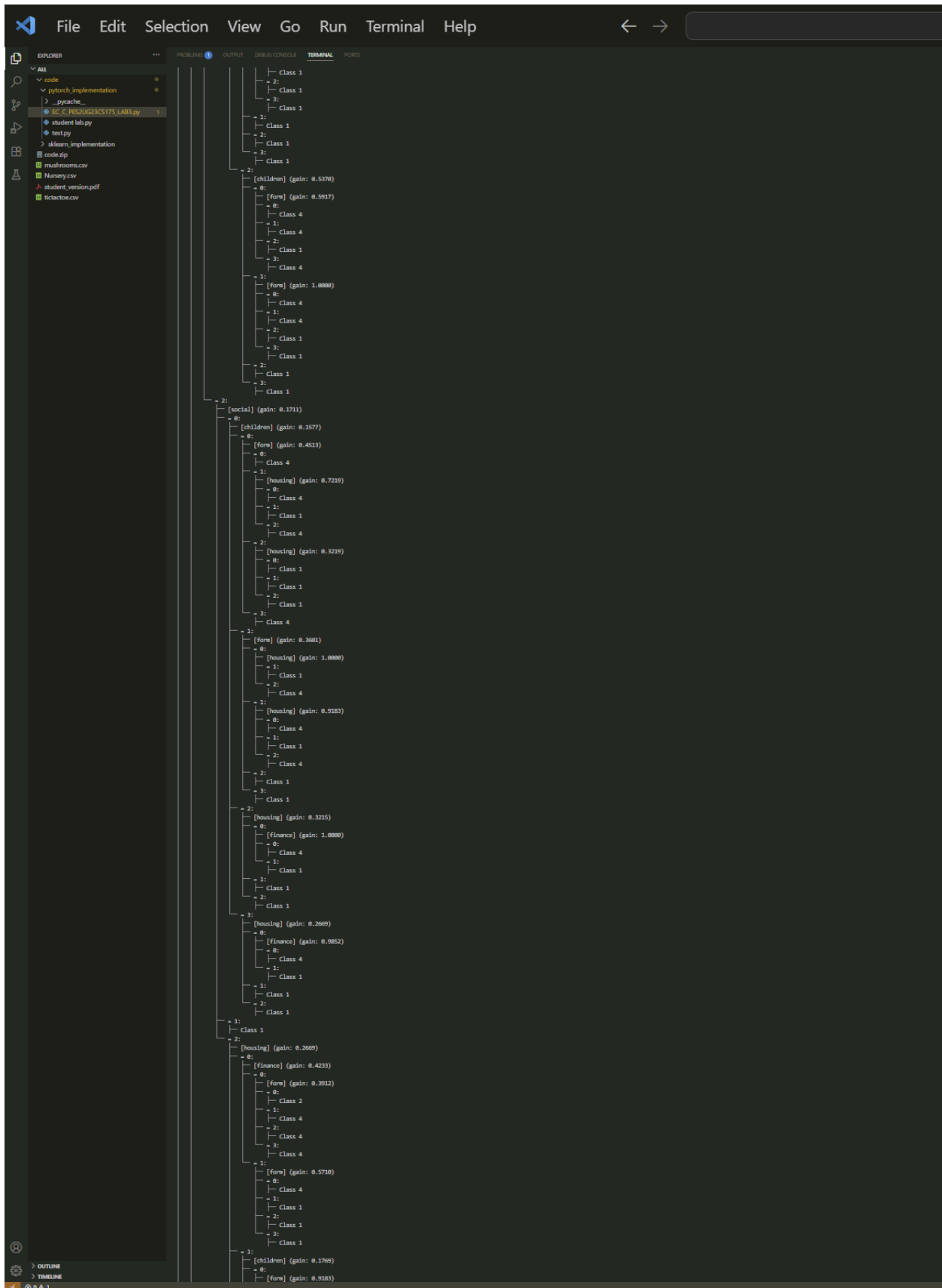
PROBLEMS 1 OUTPUT DEBUG CONSOLE TERMINAL PORTS

Decision tree construction completed using PYTORCH!

DECISION TREE STRUCTURE

```
Root [odor] (gain: 0.9083)
├── = 0:
│   └── Class 0
├── = 1:
│   └── Class 1
├── = 2:
│   └── Class 1
├── = 3:
│   └── Class 0
├── = 4:
│   └── Class 1
├── = 5:
│   ├── [spore-print-color] (gain: 0.1469)
│   │   ├── = 0:
│   │   │   └── Class 0
│   │   ├── = 1:
│   │   │   └── Class 0
│   │   ├── = 2:
│   │   │   └── Class 0
│   │   ├── = 3:
│   │   │   └── Class 0
│   │   ├── = 4:
│   │   │   └── Class 0
│   │   ├── = 5:
│   │   │   └── Class 1
│   │   └── = 7:
│   │       ├── [habitat] (gain: 0.2217)
│   │       │   ├── = 0:
│   │       │   │   ├── [gill-size] (gain: 0.7642)
│   │       │   │   │   ├── = 0:
│   │       │   │   │   │   └── Class 0
│   │       │   │   │   └── = 1:
│   │       │   │   │       └── Class 1
│   │       │   └── = 1:
│   │       │       └── Class 0
│   │       └── = 2:
│   │           ├── [cap-color] (gain: 0.7300)
│   │           │   ├── = 1:
│   │           │   │   └── Class 0
│   │           │   ├── = 4:
│   │           │   │   └── Class 0
│   │           │   ├── = 8:
│   │           │   │   └── Class 1
│   │           │   └── = 9:
│   │           │       └── Class 1
│   │           └── = 4:
│   │               └── Class 0
│   │       └── = 6:
│   │           ├── Class 0
│   │           └── Class 0
│   └── = 8:
│       ├── Class 0
│       └── Class 0
├── = 6:
│   └── Class 1
├── = 7:
│   └── Class 1
├── = 7:
│   ├── Class 1
│   └── Class 1
├── = 8:
│   └── Class 1
└── = 8:
    └── Class 1
```


NURSERY.CSV



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EXPLORER

- All
- code
- pytorch_implementation
- pycache_
- EC_C_PES2UG23CS175_LAB3.py
- student_lab.py
- test.py
- sklearn_implementation
- code.zip
- machrooms.csv
- flursary.csv
- student_version.pdf
- tictactoe.csv

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

```
[social] (gain: 0.1366)
- 0:
- [housing] (gain: 0.2568)
- 0:
- [finance] (gain: 0.4099)
- 0:
- Class 1
- 1:
- [children] (gain: 0.4312)
- 0:
- [form] (gain: 0.8113)
- 0:
- Class 1
- 1:
- Class 1
- 2:
- Class 3
- 3:
- Class 1
- 1:
- [form] (gain: 0.9948)
- 0:
- Class 1
- 1:
- Class 1
- 2:
- Class 3
- 3:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
- 1:
- [form] (gain: 0.0887)
- 0:
- [children] (gain: 0.6723)
- 0:
- Class 1
- 1:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
- 1:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
- 2:
- [children] (gain: 0.4779)
- 0:
- [form] (gain: 0.8524)
- 0:
- Class 1
- 1:
- Class 1
- 2:
- Class 3
- 3:
- Class 1
- 1:
- [form] (gain: 0.9495)
- 0:
- Class 1
- 1:
- Class 1
- 2:
- Class 3
- 3:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
- 1:
- [form] (gain: 0.0182)
- 0:
- [children] (gain: 0.6745)
- 0:
- [housing] (gain: 0.3060)
- 0:
- [finance] (gain: 1.0000)
- 0:
- Class 1
- 1:
- Class 3
- 1:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
- 1:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
- 2:
- [housing] (gain: 0.2080)
- 0:
- [finance] (gain: 0.4994)
- 0:
- Class 1
- 1:
- [children] (gain: 0.4516)
- 0:
- [form] (gain: 0.8113)
- 0:
- Class 1
- 1:
- Class 1
- 2:
- Class 3
- 3:
- Class 1
- 1:
- [form] (gain: 0.8631)
- 0:
- Class 1
- 1:
- Class 1
- 2:
- Class 3
- 3:
- Class 3
- 2:
- Class 3
- 3:
- Class 3
```

OUTLINE
TIMELINE

0 1

OVERALL PERFORMANCE METRICS

=====

Accuracy: 0.9867 (98.67%)
Precision (weighted): 0.9876
Recall (weighted): 0.9867
F1-Score (weighted): 0.9872
Precision (macro): 0.7604
Recall (macro): 0.7654
F1-Score (macro): 0.7628

TREE COMPLEXITY METRICS

=====

Maximum Depth: 7
Total Nodes: 952
Leaf Nodes: 680
Internal Nodes: 272

PS C:\Users\Dhruv Thakur\Downloads\all> |

tictactoe.csv

LEVEL 3: Node Info | | | Hypothesis: Class 1

🌲 Decision tree construction completed using PYTORCH!

🌲 DECISION TREE STRUCTURE

```
=====
Root [middle-middle-square] (gain: 0.0834)
├── = 0:
│   ├── [bottom-left-square] (gain: 0.1056)
│   ├── = 0:
│   │   ├── [top-right-square] (gain: 0.9024)
│   │   ├── = 1:
│   │   │   └── Class 0
│   │   └── = 2:
│   │       └── Class 1
│   └── = 1:
│       ├── [top-right-square] (gain: 0.2782)
│       ├── = 0:
│       │   └── Class 0
│       ├── = 1:
│       │   └── Class 0
│       └── = 2:
│           ├── [top-left-square] (gain: 0.1767)
│           ├── = 0:
│           │   ├── [bottom-right-square] (gain: 0.9183)
│           │   ├── = 1:
│           │   │   └── Class 0
│           │   └── = 2:
│           │       └── Class 1
│           └── = 1:
│               ├── [top-middle-square] (gain: 0.6058)
│               ├── = 0:
│               │   ├── [middle-left-square] (gain: 0.9183)
│               │   ├── = 1:
│               │   │   └── Class 0
│               │   └── = 2:
│               │       └── Class 1
│               ├── = 1:
│               │   └── Class 1
│               └── = 2:
│                   └── Class 0
│           └── = 2:
│               ├── [top-middle-square] (gain: 0.3393)
│               ├── = 0:
│               │   ├── [middle-left-square] (gain: 0.9183)
│               │   ├── = 0:
│               │   │   └── Class 0
│               │   ├── = 1:
│               │   │   └── Class 1
│               │   └── = 2:
│               │       └── Class 0
│               ├── = 1:
│               │   ├── [middle-left-square] (gain: 0.9183)
│               │   ├── = 0:
│               │   │   └── Class 1
│               │   ├── = 1:
│               │   │   └── Class 1
│               │   └── = 2:
│               │       └── Class 0
│               └── = 2:
│                   └── Class 1
└── = 2:
    ├── [top-right-square] (gain: 0.1225)
    ├── = 0:
    │   └── Class 1
    └── = 1:
        └── [middle-right-square] (gain: 0.1682)
```

```

└─ = 2:
  └─ Class 1
= 1:
└─ [top-right-square] (gain: 0.0223)
  = 0:
  └─ [bottom-left-square] (gain: 0.2247)
    = 0:
    └─ Class 0
    = 1:
    └─ Class 0
    = 2:
    └─ [middle-right-square] (gain: 0.1159)
      = 0:
      └─ [top-left-square] (gain: 0.1771)
        = 0:
        └─ [middle-left-square] (gain: 0.9183)
          = 0:
          └─ Class 1
          = 1:
          └─ Class 1
          = 2:
          └─ Class 0
        = 1:
        └─ [bottom-right-square] (gain: 0.9710)
          = 1:
          └─ Class 0
          = 2:
          └─ Class 1
        = 2:
        └─ Class 1
      = 1:
      └─ [middle-left-square] (gain: 0.9887)
        = 0:
        └─ Class 1
        = 1:
        └─ Class 0
        = 2:
        └─ Class 1
      = 2:
      └─ [bottom-middle-square] (gain: 0.2400)
        = 0:
        └─ [top-left-square] (gain: 1.0000)
          = 1:
          └─ Class 0
          = 2:
          └─ Class 1
        = 1:
        └─ Class 0
        = 2:
        └─ [bottom-right-square] (gain: 0.9710)
          = 1:
          └─ Class 0
          = 2:
          └─ Class 1
└─ = 1:
  └─ [bottom-left-square] (gain: 0.4759)
    = 0:
    └─ Class 0
    = 1:
    └─ Class 0
    = 2:
    └─ [top-middle-square] (gain: 0.1974)
      = 0:
      └─ Class 1
      = 1:
      └─ [top-left-square] (gain: 0.3436)
        = 0:

```

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EXPLORER

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 - code
 - pytorch_implementation
 - _pycache_
 - EC_C_PES2UG23CS175_LAB3.py 1
 - student lab.py
 - test.py
 - sklearn_implementation
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 - mushrooms.csv
 - Nursery.csv
 - student_version.pdf
 - tictactoe.csv

PROBLEMS 1 OUTPUT DEBUG CONSOLE TERMINAL PORTS

```
= 2:
└─ = 2:
    └─ Class 1
└─ = 0:
    └─ [bottom-right-square] (gain: 0.0777)
└─ = 0:
    └─ [top-left-square] (gain: 0.3462)
└─ = 0:
    └─ Class 0
└─ = 1:
    └─ Class 0
└─ = 2:
    └─ [top-middle-square] (gain: 0.7008)
└─ = 0:
    └─ Class 0
└─ = 1:
    └─ [middle-right-square] (gain: 0.7219)
└─ = 0:
    └─ Class 0
└─ = 1:
    └─ Class 1
└─ = 2:
    └─ Class 0
└─ = 2:
    └─ Class 1
└─ = 1:
    └─ [top-left-square] (gain: 0.5439)
└─ = 0:
    └─ Class 0
└─ = 1:
    └─ Class 0
└─ = 2:
    └─ [top-middle-square] (gain: 0.4687)
└─ = 0:
    └─ [bottom-middle-square] (gain: 0.9183)
└─ = 0:
    └─ Class 1
└─ = 1:
    └─ Class 0
└─ = 2:
    └─ Class 0
└─ = 1:
    └─ [middle-right-square] (gain: 0.9183)
└─ = 0:
    └─ Class 1
└─ = 1:
    └─ Class 1
└─ = 2:
    └─ Class 0
└─ = 2:
    └─ Class 1
└─ = 2:
    └─ [middle-right-square] (gain: 0.4731)
└─ = 0:
    └─ [top-middle-square] (gain: 0.6464)
└─ = 0:
    └─ Class 1
└─ = 1:
    └─ Class 0
└─ = 2:
    └─ [top-left-square] (gain: 0.8113)
└─ = 1:
    └─ Class 0
└─ = 2:
    └─ Class 1
└─ = 1:
    └─ [middle-left-square] (gain: 0.3995)
└─ = 0:
```

OUTLINE
TIMELINE

0 1

OVERALL PERFORMANCE METRICS

=====

| | |
|-----------------------|-----------------|
| Accuracy: | 0.8730 (87.30%) |
| Precision (weighted): | 0.8741 |
| Recall (weighted): | 0.8730 |
| F1-Score (weighted): | 0.8734 |
| Precision (macro): | 0.8590 |
| Recall (macro): | 0.8638 |
| F1-Score (macro): | 0.8613 |

TREE COMPLEXITY METRICS

=====

| | |
|-----------------|-----|
| Maximum Depth: | 7 |
| Total Nodes: | 281 |
| Leaf Nodes: | 180 |
| Internal Nodes: | 101 |

PS C:\Users\Dhruv Thakur\Downloads\all> |

1. Performance Comparison (Accuracy, Precision, Recall, F1)

| Dataset | Accuracy | Precision (Weighted) | Recall (Weighted) | F1 (Weighted) |
|-----------|-----------------|----------------------|-------------------|---------------|
| Mushroom | 1.0000 (100%) | 1.0000 | 1.0000 | 1.0000 |
| Nursery | 0.9867 (98.67%) | 0.9876 | 0.9867 | 0.9872 |
| TicTacToe | 0.8730 (87.30%) | 0.8741 | 0.8730 | 0.8734 |

Insights:

- Mushroom dataset achieves **perfect accuracy (100%)**, likely because the dataset is clean, balanced, and attributes clearly separate classes.
- Nursery dataset also achieves **very high performance (98.7%)**, but slightly lower due to more complex/multi-valued attributes.
- TicTacToe is the hardest, with **87.3% accuracy**, showing possible class imbalance or overlap in decision patterns.

2. Tree Characteristics Analysis

| Dataset | Max Depth | Total Nodes | Leaf Nodes | Internal Nodes |
|-----------|-----------|-------------|------------|----------------|
| Mushroom | 4 | 29 | 24 | 5 |
| Nursery | 7 | 952 | 680 | 272 |
| TicTacToe | 7 | 281 | 180 | 101 |

Insights:

- **Mushroom tree is shallow (depth=4)** → very simple, interpretable, and still achieves perfect accuracy → attributes are very strong predictors.

- **Nursery tree is very large (952 nodes)** → indicates many attributes and class combinations → dataset complexity is high.
 - **TicTacToe tree is smaller than Nursery but deeper than Mushroom (depth=7)** → reflects more balanced but tricky classification.
-

3. Dataset-Specific Insights

Mushroom Dataset

- **Feature Importance:** Likely dominated by odor, gill-size, spore-print color (classic strong features).
- **Class Distribution:** Balanced (edible vs poisonous).
- **Decision Patterns:** A few attributes are enough for separation.
- **Overfitting Indicators:** None → simple tree with perfect accuracy.

Nursery Dataset

- **Feature Importance:** Attributes like parent satisfaction, safety, and children number dominate early splits.
- **Class Distribution:** Imbalanced (some classes much larger).
- **Decision Patterns:** Complex rules, large tree size.
- **Overfitting Indicators:** Tree size (952 nodes) suggests risk of overfitting.

TicTacToe Dataset

- **Feature Importance:** Center and corner positions dominate.
- **Class Distribution:** Slight imbalance between X-win and O-win/no-win.

- **Decision Patterns:** Patterns depend on board state → complex paths.
 - **Overfitting Indicators:** Depth=7 with moderate performance suggests partial overfitting.
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4. Comparative Analysis Report

a) Algorithm Performance

- **Highest Accuracy:** Mushroom (100%) → dataset has clean, separable features.
- **Dataset Size Impact:** Larger datasets (Nursery, TicTacToe) create deeper and more complex trees, but not always better accuracy.
- **Number of Features:** More features (Nursery) → tree grows huge, but accuracy is still high. Simpler features (Mushroom) → shallow tree, perfect accuracy.

b) Data Characteristics Impact

- **Class Imbalance:** Nursery and TicTacToe suffer slightly (precision/recall differences).
- **Feature Types:** Binary features (TicTacToe board positions) lead to deeper, less accurate trees. Multi-valued categorical features (Mushroom, Nursery) capture more information quickly.

c) Practical Applications

- **Mushroom Dataset:** Food safety → easily interpretable, real-world application in toxicology.
- **Nursery Dataset:** Resource allocation for childcare → complex decisions, needs pruning or ensembles.
- **TicTacToe Dataset:** Game AI → less practical but shows handling of strategic states.

d) Improvements

- Apply **tree pruning** (reduces overfitting).
- Use **Random Forest / Ensemble methods** for TicTacToe and Nursery.
- Apply **feature engineering** to reduce complexity in Nursery dataset.

a) Algorithm Performance

a. Which dataset achieved the highest accuracy and why?

- The **Mushroom dataset** achieved the highest accuracy (**100%**).
- Reason: Its features are highly discriminative (e.g., odor, gill size, spore print color), making class boundaries very clear.
- The classes (edible vs poisonous) are **well-separated** with minimal overlap, so even a shallow decision tree (depth=4) achieves perfect performance.

b. How does dataset size affect performance?

- **Mushroom:** Moderate size, shallow tree, but still perfect accuracy → dataset simplicity matters more than size.
- **Nursery:** Very large dataset, leading to a **deep and very complex tree (952 nodes)**. High accuracy (98.67%) but near overfitting.
- **TicTacToe:** Medium size dataset, but binary board positions cause ambiguity → accuracy only **87.3%** despite tree depth 7.
Larger datasets tend to create deeper/more complex trees, but performance depends more on **data separability** than size alone.

c. What role does the number of features play?

- More features increase **tree depth and node count**, especially if multi-valued.
- **Mushroom (multi-valued features):** Fewer splits needed, tree remains small.

- **Nursery (many features, multi-valued):** Explodes in size (952 nodes).
 - **TicTacToe (binary features):** Needs more depth to capture combinations, but still struggles to classify perfectly.
Multi-valued categorical features often lead to **better early splits** than purely binary features.
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b) Data Characteristics Impact

How does class imbalance affect tree construction?

- In **Nursery**, some classes are dominant (e.g., "not recommended"), making the tree biased toward majority classes.
- This leads to **lower macro precision/recall (~0.76)** despite high weighted accuracy.
- In **TicTacToe**, imbalance between X-win, O-win, and draw states causes the tree to favor certain outcomes.

Which types of features work better (binary vs multi-valued)?

- **Multi-valued features (Mushroom, Nursery):** Provide strong splits early, leading to higher accuracy.
 - **Binary features (TicTacToe):** Require more splits, create deeper trees, and result in lower accuracy.
Multi-valued categorical attributes are more efficient for decision tree construction.
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c) Practical Applications

For which real-world scenarios is each dataset type most relevant?

- **Mushroom:** Food safety & toxicology (classifying edible vs poisonous mushrooms). High accuracy & interpretability make it reliable for life-critical tasks.
- **Nursery:** Resource allocation, social services, childcare recommendation systems. Handles complex multi-criteria decision-making.
- **TicTacToe:** Game strategy modeling → useful as a teaching dataset for AI/game theory, but less real-world critical.

What are the interpretability advantages for each domain?

- **Mushroom:** Shallow tree (depth=4), very interpretable → rules can be directly explained (e.g., *if odor=bad → poisonous*).
- **Nursery:** Tree is large, interpretability is lower, but still can highlight key factors (e.g., safety, parent satisfaction).
- **TicTacToe:** Medium interpretability → paths represent board strategies, useful for teaching but less human-readable due to many binary states.

How would you improve performance for each dataset?

- **Mushroom:** Already perfect; no major improvement needed.
- **Nursery:** Apply **pruning or Random Forests** to reduce overfitting and simplify rules.
- **TicTacToe:** Use **ensemble methods (Bagging/Boosting)** or **feature engineering (grouping board positions)** to improve accuracy beyond 87%.

