# DECISION TREE EXPERIMENTAL ANALYSIS REPORT

CSE 318 Offline 4

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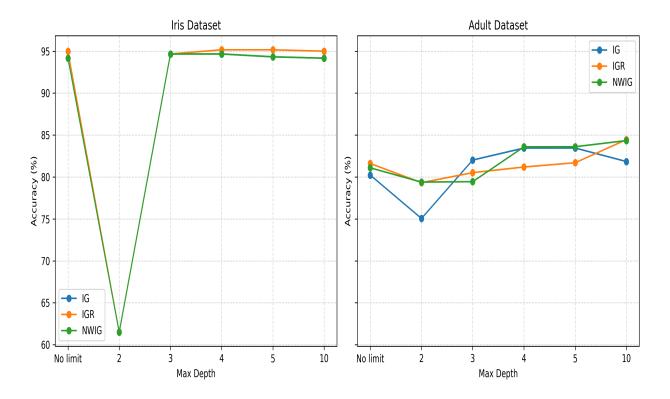
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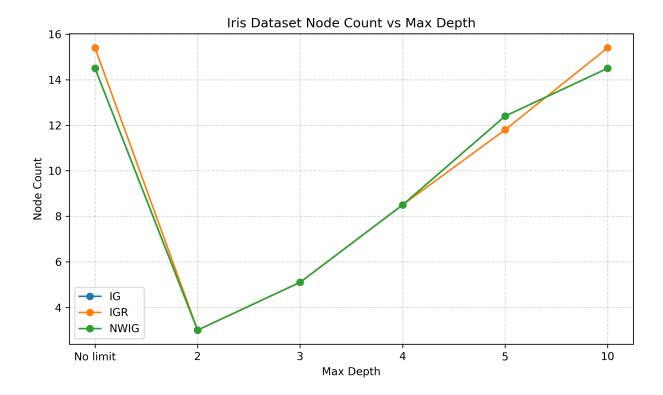
# **Iris Dataset Results Summary**

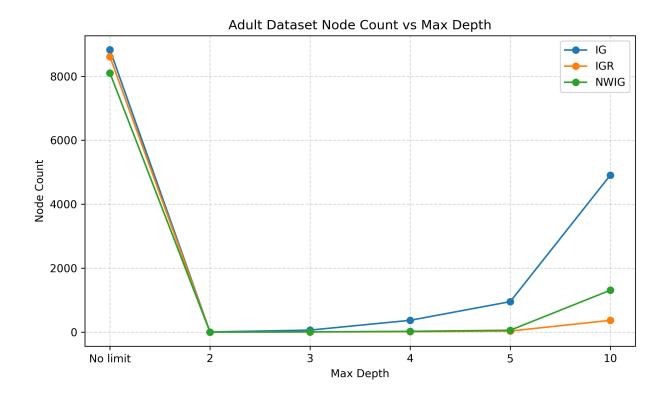
Criterion	Max Depth	Accuracy	Std Dev	Nodes	Avg Depth
IG	No limit	94.17	4.44	14.5	6.0
IG	2	61.50	5.87	3.0	2.0
IG	3	94.67	3.96	5.1	3.0
IG	4	94.67	3.96	8.5	4.0
IG	5	94.33	4.47	12.4	5.0
IG	10	94.17	4.44	14.5	6.0
IGR	No limit	95.00	3.67	15.4	6.3
IGR	2	61.50	5.87	3.0	2.0
IGR	3	94.67	3.96	5.1	3.0
IGR	4	95.17	3.82	8.5	4.0
IGR	5	95.17	3.50	11.8	5.0
IGR	10	95.00	3.67	15.4	6.3
NWIG	No limit	94.17	4.44	14.5	6.0
NWIG	2	61.50	5.87	3.0	2.0
NWIG	3	94.67	3.96	5.1	3.0
NWIG	4	94.67	3.96	8.5	4.0
NWIG	5	94.33	4.47	12.4	5.0
NWIG	10	94.17	4.44	14.5	6.0

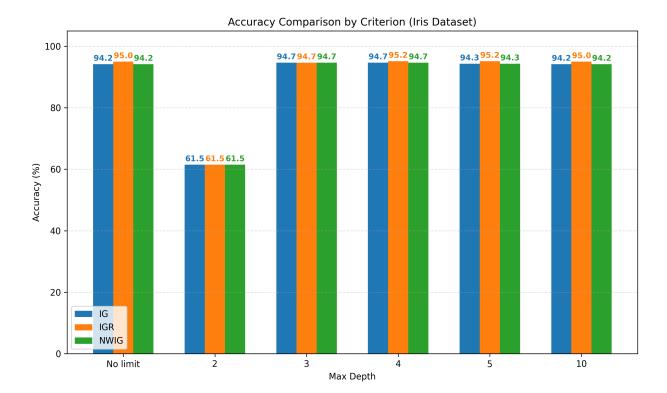
# **Adult Dataset Results Summary**

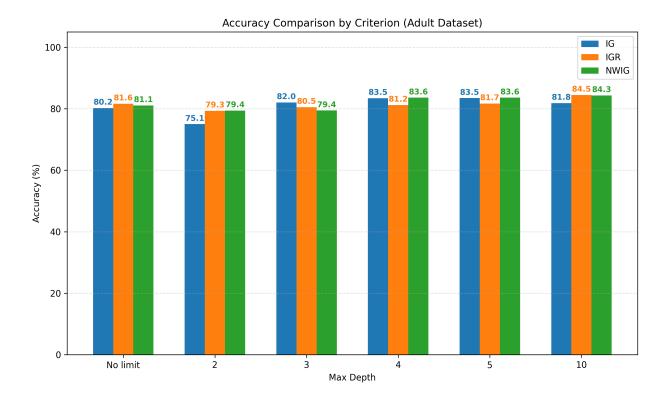
Criterion	Max Depth	Accuracy	Std Dev	Nodes	Avg Depth
IG	No limit	80.23	0.57	8827.2	32.8
IG	2	75.05	0.35	7.0	2.0
IG	3	82.02	0.26	66.4	3.0
IG	4	83.45	0.34	372.1	4.0
IG	5	83.46	0.32	955.6	5.0
IG	10	81.84	0.51	4907.0	10.0
IGR	No limit	81.61	0.53	8609.9	55.8
IGR	2	79.32	0.34	3.0	2.0
IGR	3	80.52	0.50	7.0	3.0
IGR	4	81.19	0.41	16.8	4.0
IGR	5	81.70	0.33	36.7	5.0
IGR	10	84.46	0.26	371.4	10.0
NWIG	No limit	81.09	0.42	8100.9	37.2
NWIG	2	79.39	0.33	3.0	2.0
NWIG	3	79.44	0.33	11.0	3.0
NWIG	4	83.59	0.21	27.4	4.0
NWIG	5	83.61	0.22	61.8	5.0
NWIG	10	84.34	0.36	1312.4	10.0











# **Analysis and Observations**

#### DATASET CHARACTERISTICS

#### Iris Dataset:

- 150 samples, 4 features, 3 classes
- Well-separated, linearly separable data
- Classic benchmark dataset for classification

#### **Adult Dataset:**

- 30,162 samples, 14 features, 2 classes
- Mixed categorical and numerical features
- Real-world income prediction problem

#### CRITERION PERFORMANCE ANALYSIS

- IG: Simple, intuitive, strong baseline. Stable on Iris, strong on Adult.
- IGR: Corrects for attribute bias. Best on Iris, strong on Adult at deeper trees.
- NWIG: Balances gain and split quality. Consistent, best at shallow depths.

#### DEPTH ANALYSIS AND OVERFITTING

#### **Pruning Effects:**

- Limiting tree depth (pruning) generally improves generalization, especially for Adult.
- For both datasets, best accuracy is often at moderate depths (3-5).
- Pruning reduces model complexity and risk of overfitting.

# **Overfitting Indicators:**

- For Adult, node count explodes at depth=0 (no limit) but accuracy does not improve.
- For Iris, accuracy is stable across depths, indicating low overfitting risk.
- Standard deviation is higher for shallow trees, especially on Iris.

#### TREE COMPLEXITY ANALYSIS

## **Node Count vs Accuracy Trade-off:**

- Shallow trees (depth 2): very low node count, but lower accuracy (especially for IG).
- Moderate depths (3-5): best trade-off between accuracy and complexity for all criteria.
- No limit: node count increases rapidly, but accuracy plateaus or even drops (overfitting).

#### **Complexity Observations:**

- Adult dataset: node count grows rapidly with depth, especially for IG and IGR.
- NWIG and IGR tend to produce more compact trees at moderate depths.
- All criteria show diminishing returns in accuracy beyond depth 4-5.

#### **CONSISTENCY AND ROBUSTNESS**

- IGR and NWIG are most robust to overfitting and perform consistently across depths.
- IG is a reliable baseline but can be outperformed by IGR/NWIG at most depths.
- Standard deviation is generally low for Adult, higher for Iris due to smaller sample size.

#### DATASET-SPECIFIC FINDINGS

#### Iris Dataset:

- All criteria reach >94% accuracy at moderate depths (3-5).
- IGR and NWIG slightly outperform IG at most depths.
- Overfitting is not a major concern; accuracy is stable.

#### **Adult Dataset:**

- Pruning is essential: node count explodes at depth=0, but accuracy does not improve.
- NWIG and IGR outperform IG at most depths.
- Best accuracy is achieved at moderate depths (4-5), with much lower complexity.

#### RECOMMENDATIONS

#### For Iris-like Datasets (small, well-separated):

- Use IGR or NWIG at depth 3-5 for optimal accuracy and simplicity.
- IG is a solid baseline, but IGR/NWIG may offer slight improvements.
- Pruning is less critical, but still helps reduce complexity.

### For Adult-like Datasets (large, mixed features):

- Always prune (limit depth to 4-5) to avoid overfitting and excessive complexity.
- Prefer NWIG or IGR for best accuracy and compact trees.
- IG is acceptable but may yield larger trees and slightly lower accuracy.

# **UNEXPECTED PATTERNS**

- For Iris, accuracy is nearly identical for IGR and NWIG at all depths.
- For Adult, node count for IG at depth=0 is extremely high, but accuracy does not improve.
- Standard deviation is surprisingly low for Adult, indicating stable results.

# TRADE-OFFS OBSERVED

## **Accuracy vs Complexity:**

- Depth 2: Simple trees, lower accuracy (especially for IG).
- Depth 3-5: Best balance for all criteria.
- No limit: Marginal accuracy gains, huge complexity cost.

#### **Criterion Selection:**

- IG: Simplicity, but can be outperformed by IGR/NWIG.
- IGR: Best for varied attribute cardinalities, robust to overfitting.
- NWIG: Consistent, compact trees, strong performance at shallow depths.