



# **BoolMinGeo**

## 4D Minimization Performance

### 9-10 Variable Boolean Functions

Total Tests: 106

Date: 2026-01-07

EXPERIMENTAL SETUP & CONFIGURATION

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STUDY INFORMATION

Study Type:	Performance Characterization
Scope:	9-16 variable Boolean functions
Total Tests:	106
Date:	2026-01-07

SYSTEM CONFIGURATION

Platform:	Windows-11-10.0.26200-SP0
Processor:	Intel64 Family 6 Model 142 Stepping 12, GenuineIntel
Python:	3.12.10

SOFTWARE VERSIONS

NumPy:	2.3.4
SciPy:	1.16.3
Matplotlib:	3.10.7

EXPERIMENTAL PARAMETERS

Random Seed:	42
Variable Range:	9-10
Tests per Distribution:	10

TEST DISTRIBUTIONS

• Sparse:	20% ones, 5% don't-cares
• Dense:	70% ones, 5% don't-cares
• Balanced:	50% ones, 10% don't-cares
• Minimal DC:	45% ones, 2% don't-cares
• Heavy DC:	30% ones, 30% don't-cares
• Edge cases:	all-zeros, all-ones, all-dc

METRICS COLLECTED

• Execution time (seconds)
• Memory consumption (MB)
• Peak memory usage (MB)
• Solution complexity (literal count, term count)
• Time per truth table entry (ms)
• Memory per truth table entry (KB)

METHODOLOGY

1. Random Boolean functions generated per distribution
2. BoolMinGeo 4D minimization executed (SOP form)
3. Execution time measured using perf_counter
4. Memory tracked using tracemalloc + psutil
5. Results aggregated and analyzed statistically
6. Exponential models fitted to scaling data
7. Extrapolations computed for larger problems

NOTE ON SYMPY COMPARISON

This is a performance-only study. SymPy comparison is omitted for 9-10 variables due to computational infeasibility. See verify_sympy_failure.py for detailed justification.
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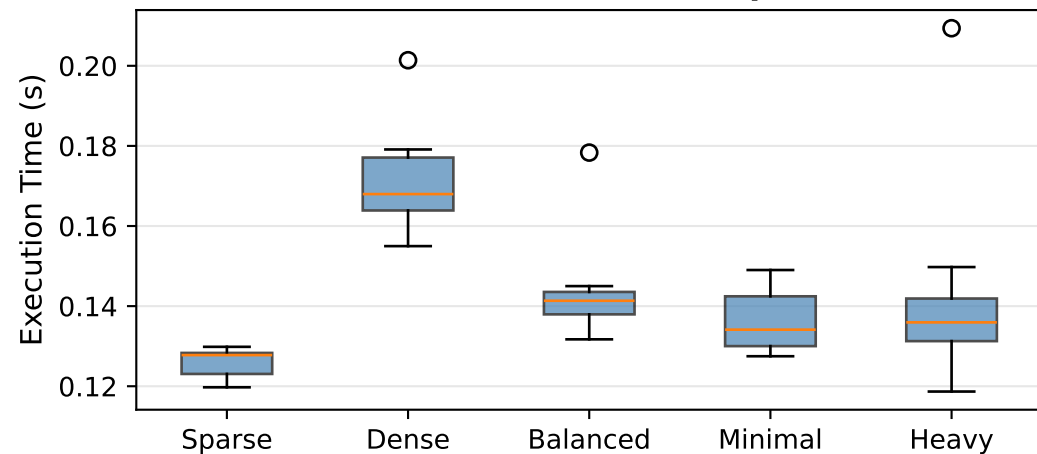
REPRODUCIBILITY

To reproduce this experiment:
1. Set random seed: random.seed(42)
2. Run with identical system configuration
3. Use same library versions as documented above
4. Execute: python benchmark_test4D.py

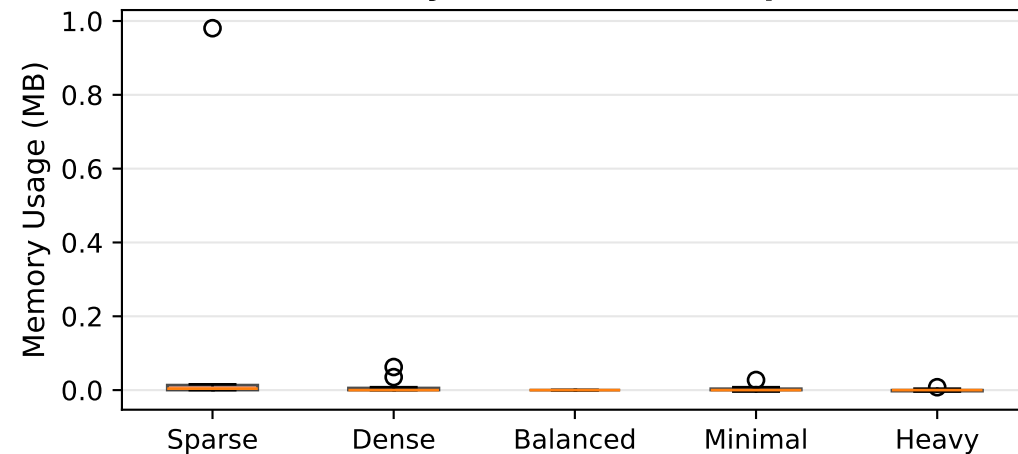
# 9-Variable K-Map: Distribution Performance Analysis

## Truth Table Size: $2^9 = 512$ entries

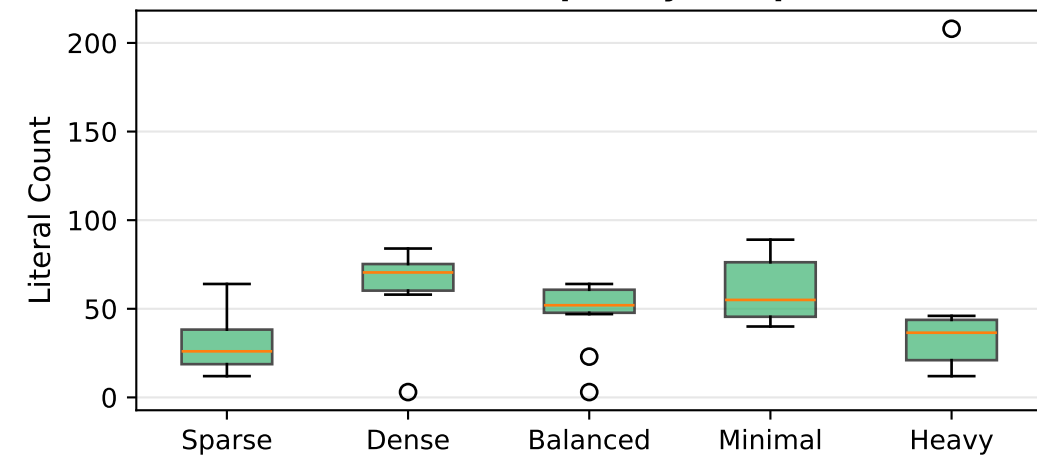
### A) Time Distribution Comparison



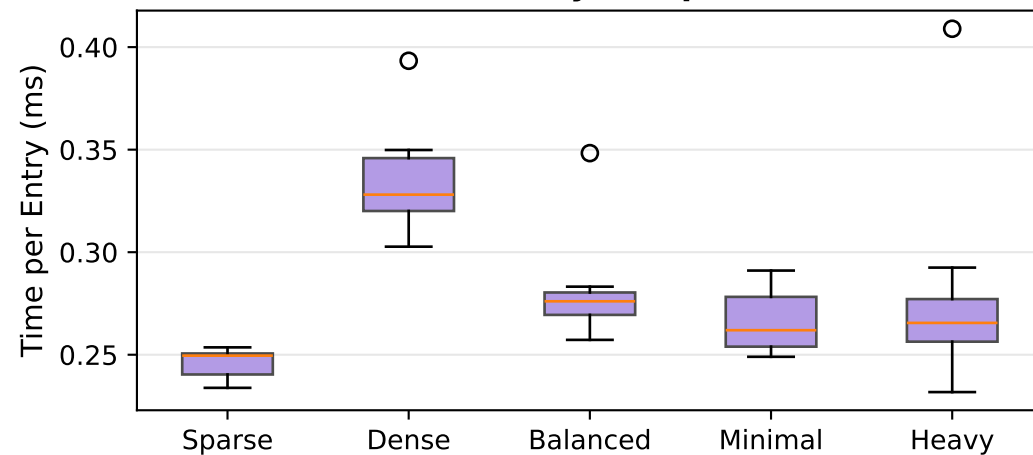
### B) Memory Distribution Comparison



### C) Solution Complexity Comparison



### D) Efficiency Comparison



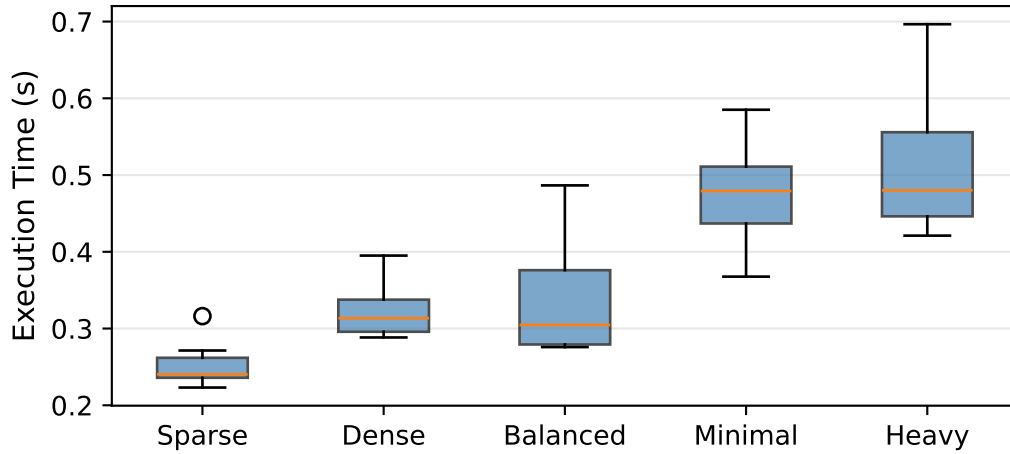
### E) Statistical Summary

Distribution	N	Mean Time (s)	Std Time	Mean Mem (MB)	Mean Lits	Mean Terms
Sparse (20% 1s)	10	0.1260	0.0036	0.10	31.3	7.4
Dense (70% 1s)	10	0.1710	0.0126	0.01	63.4	18.9
Balanced (50% 1s)	10	0.1435	0.0123	0.00	47.4	13.1
Minimal DC (2%)	10	0.1362	0.0073	0.00	60.3	15.5
Heavy DC (30%)	10	0.1422	0.0238	0.00	48.8	12.1

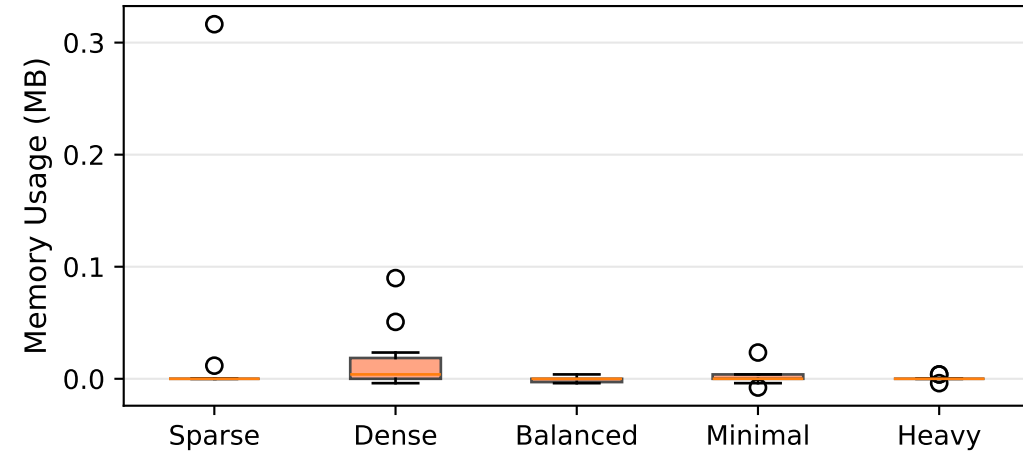
# 10-Variable K-Map: Distribution Performance Analysis

## Truth Table Size: $2^{10} = 1,024$ entries

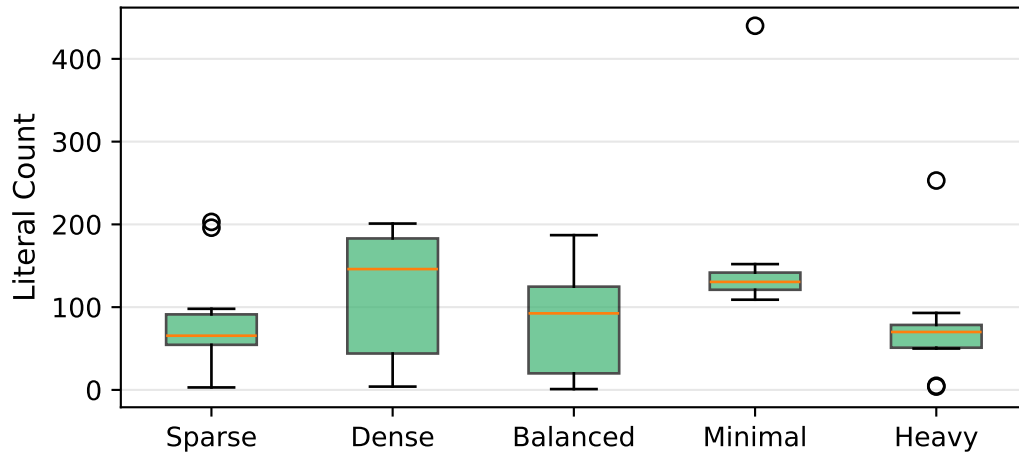
### A) Time Distribution Comparison



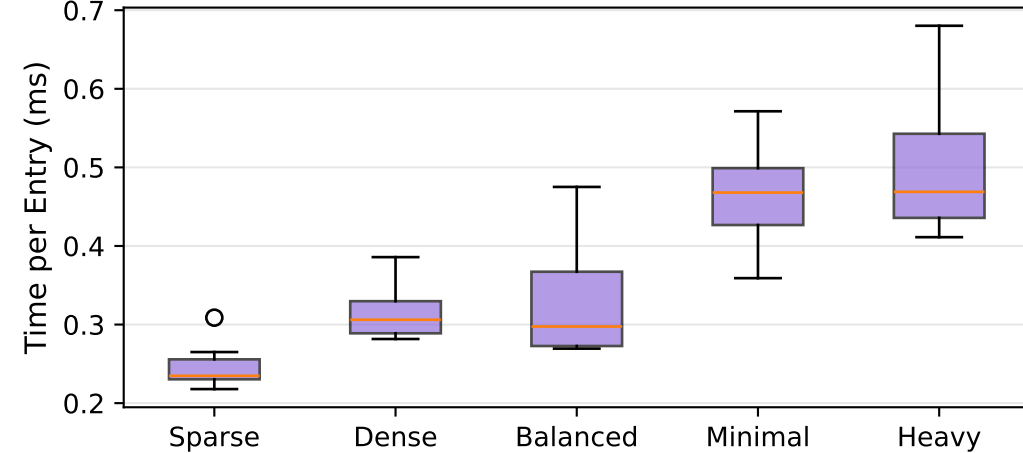
### B) Memory Distribution Comparison



### C) Solution Complexity Comparison



### D) Efficiency Comparison

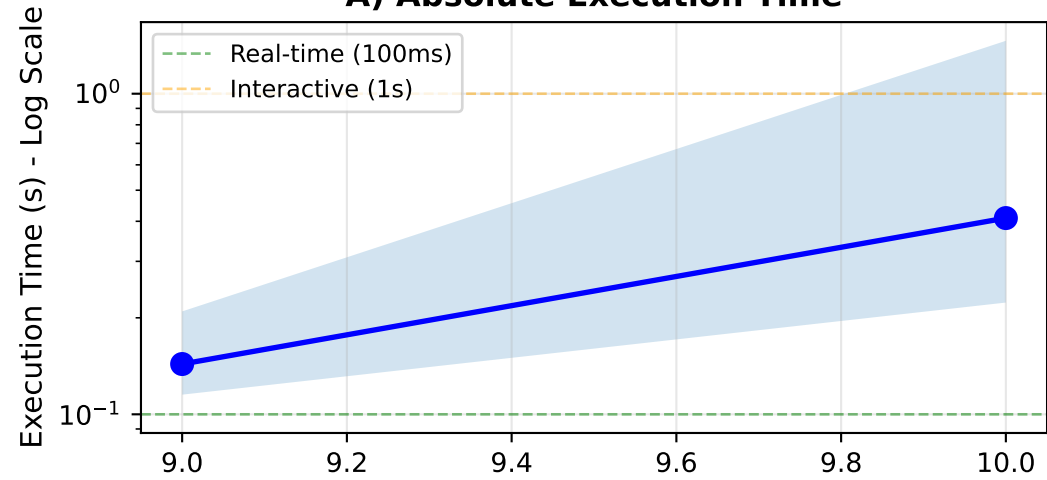


### E) Statistical Summary

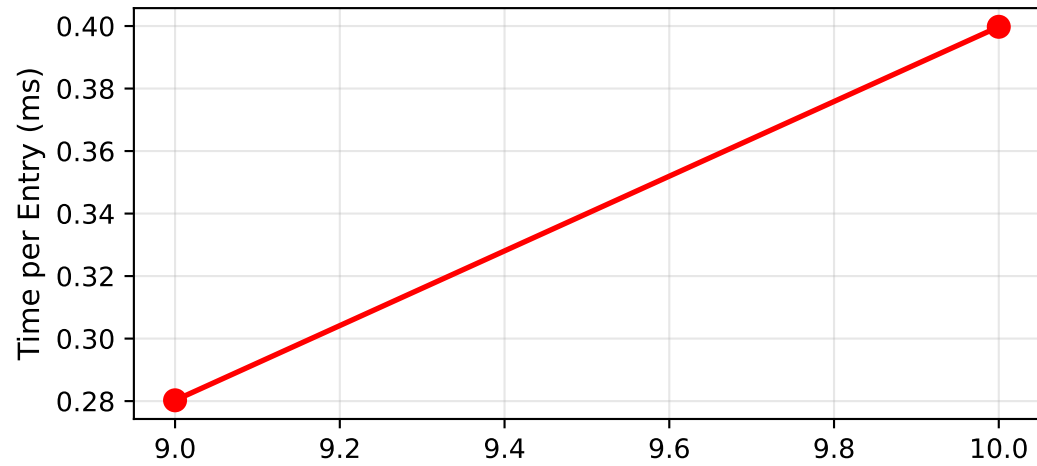
Distribution	N	Mean Time (s)	Std Time	Mean Mem (MB)	Mean Lits	Mean Terms
Sparse (20% 1s)	10	0.2510	0.0259	0.03	85.6	18.6
Dense (70% 1s)	10	0.3210	0.0319	0.02	119.8	30.4
Balanced (50% 1s)	10	0.3358	0.0740	-0.00	87.6	21.0
Minimal DC (2%)	10	0.4772	0.0574	0.00	160.4	34.1
Heavy DC (30%)	10	0.5098	0.0865	0.00	75.5	17.5

# BoolMinGeo 4D Minimization Performance Characterization (9-10 Variables)

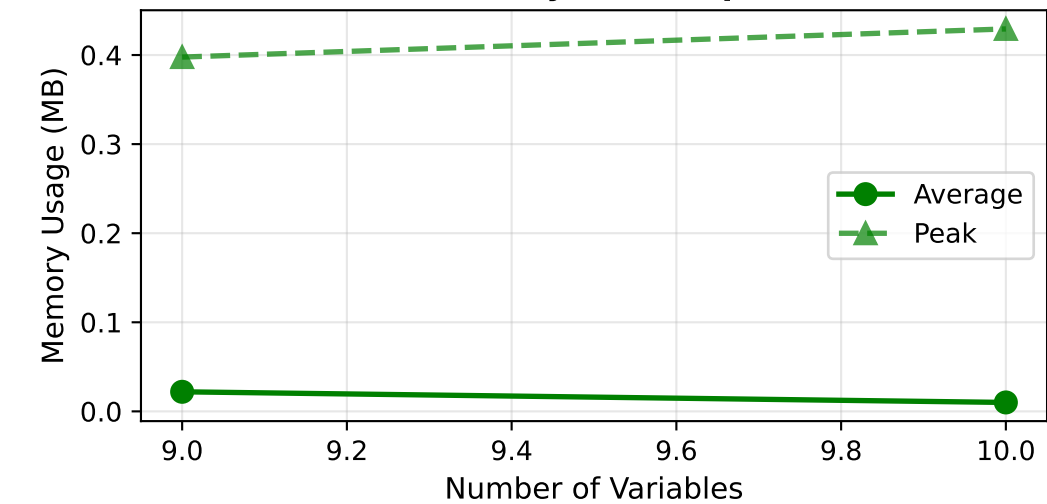
**A) Absolute Execution Time**



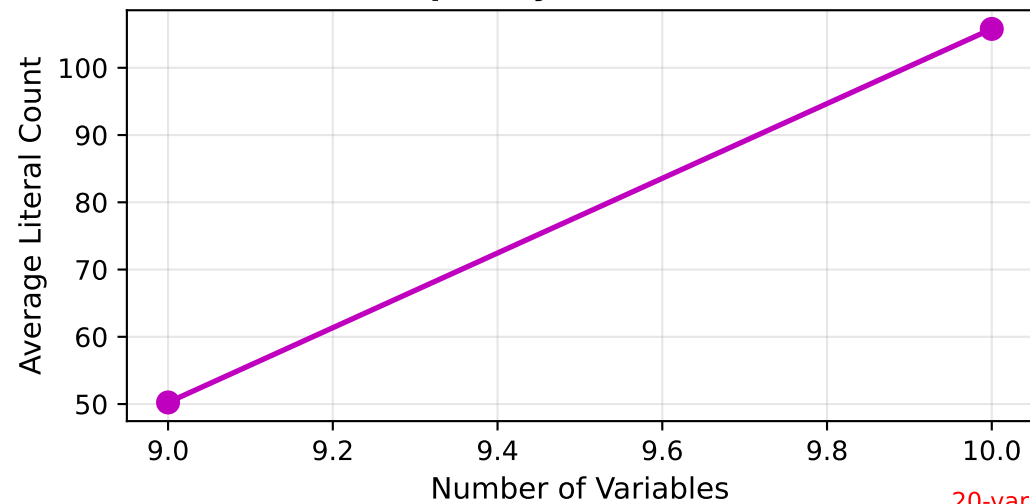
**B) Efficiency (Time per Truth Table Entry)**



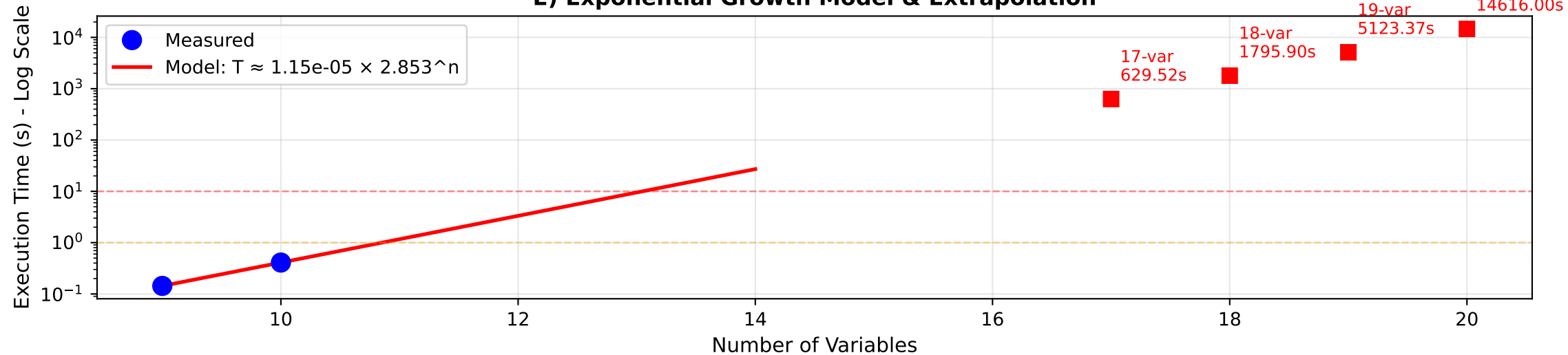
**C) Memory Consumption**



**D) Solution Complexity (Non-constant Functions)**

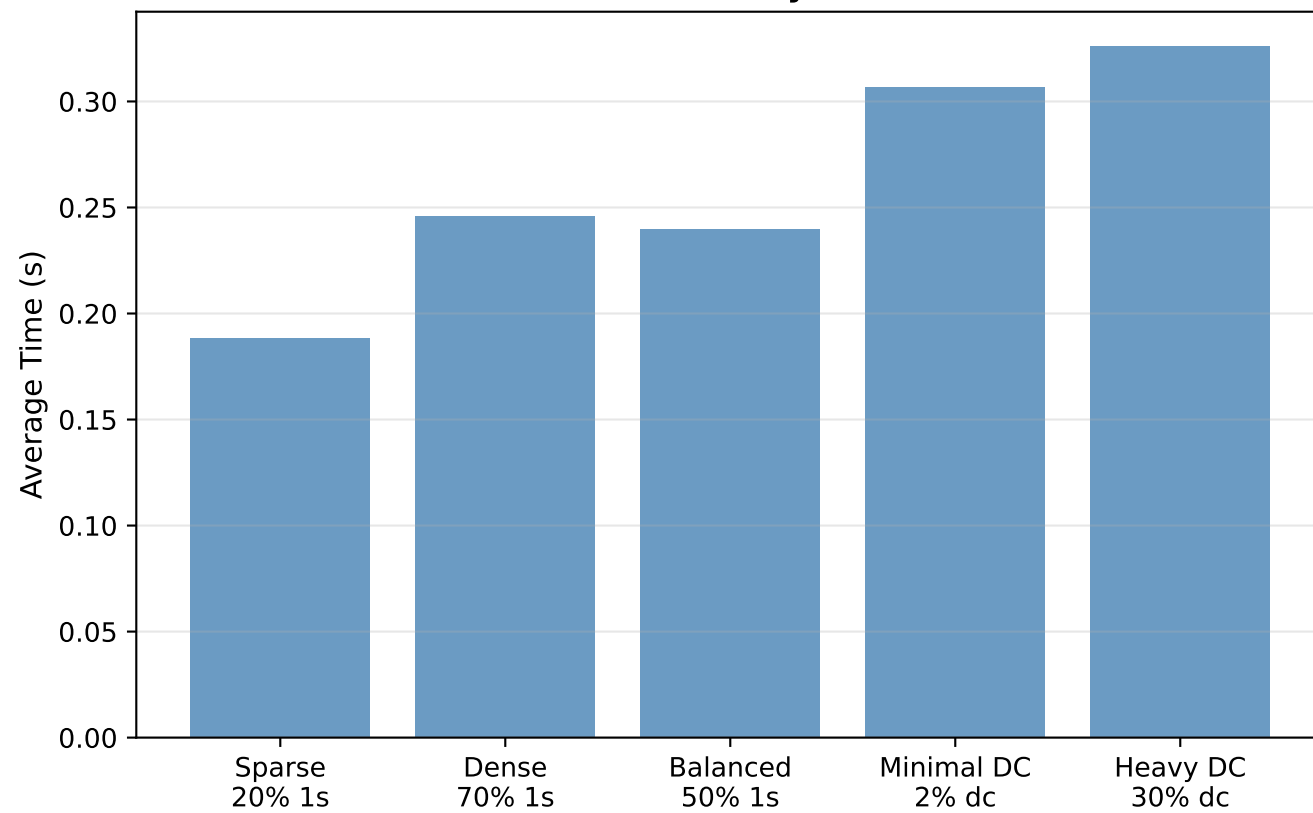


**E) Exponential Growth Model & Extrapolation**

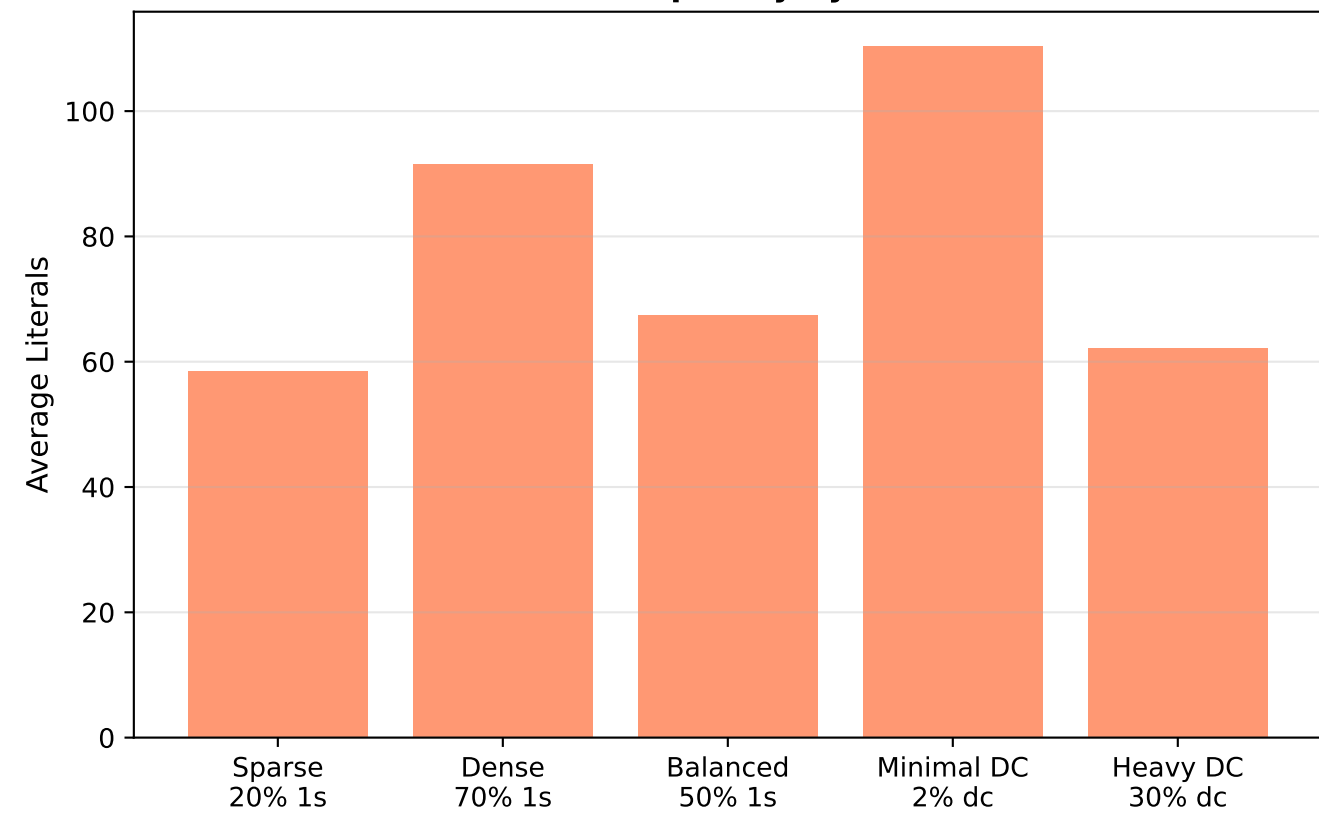


# Distribution Sensitivity Analysis

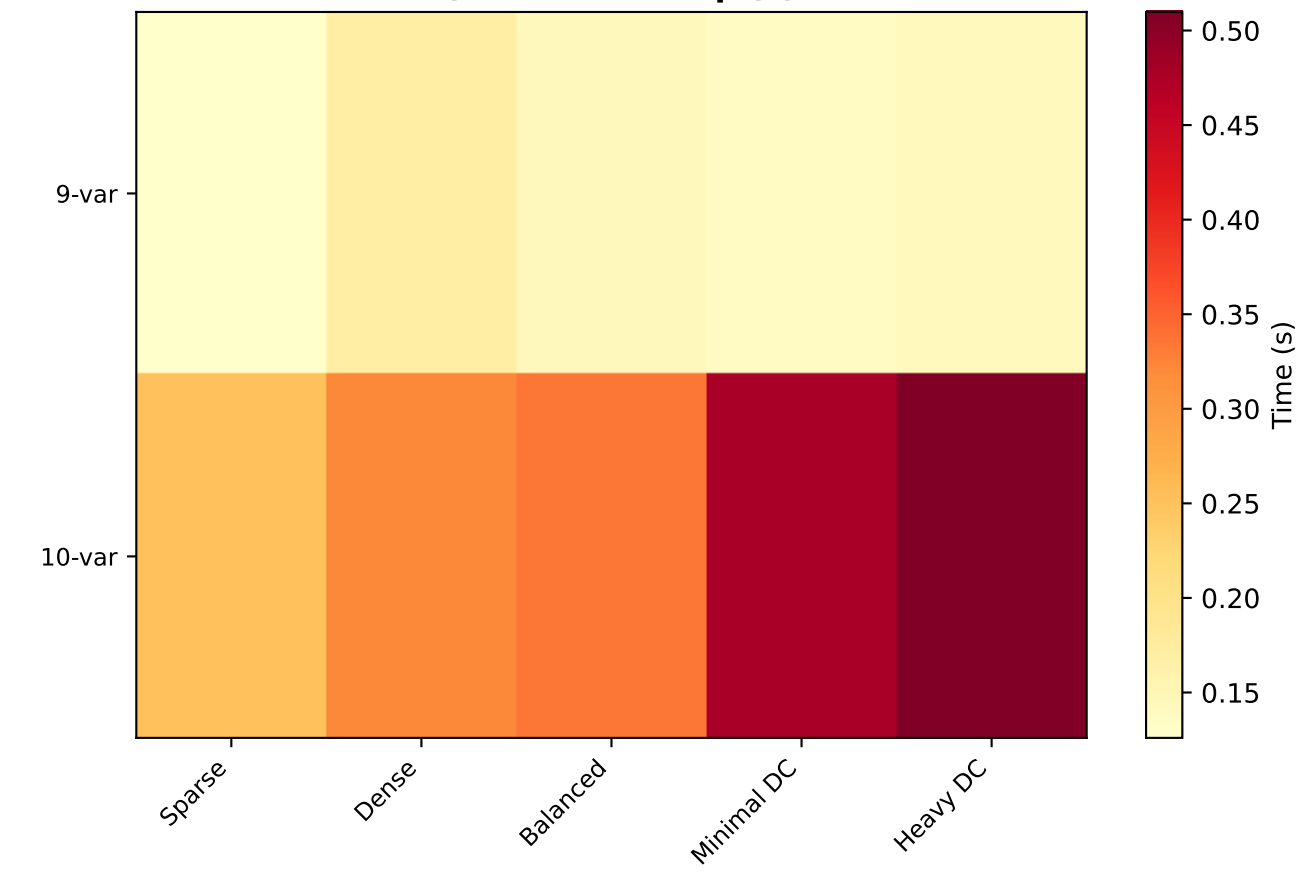
## A) Execution Time by Distribution



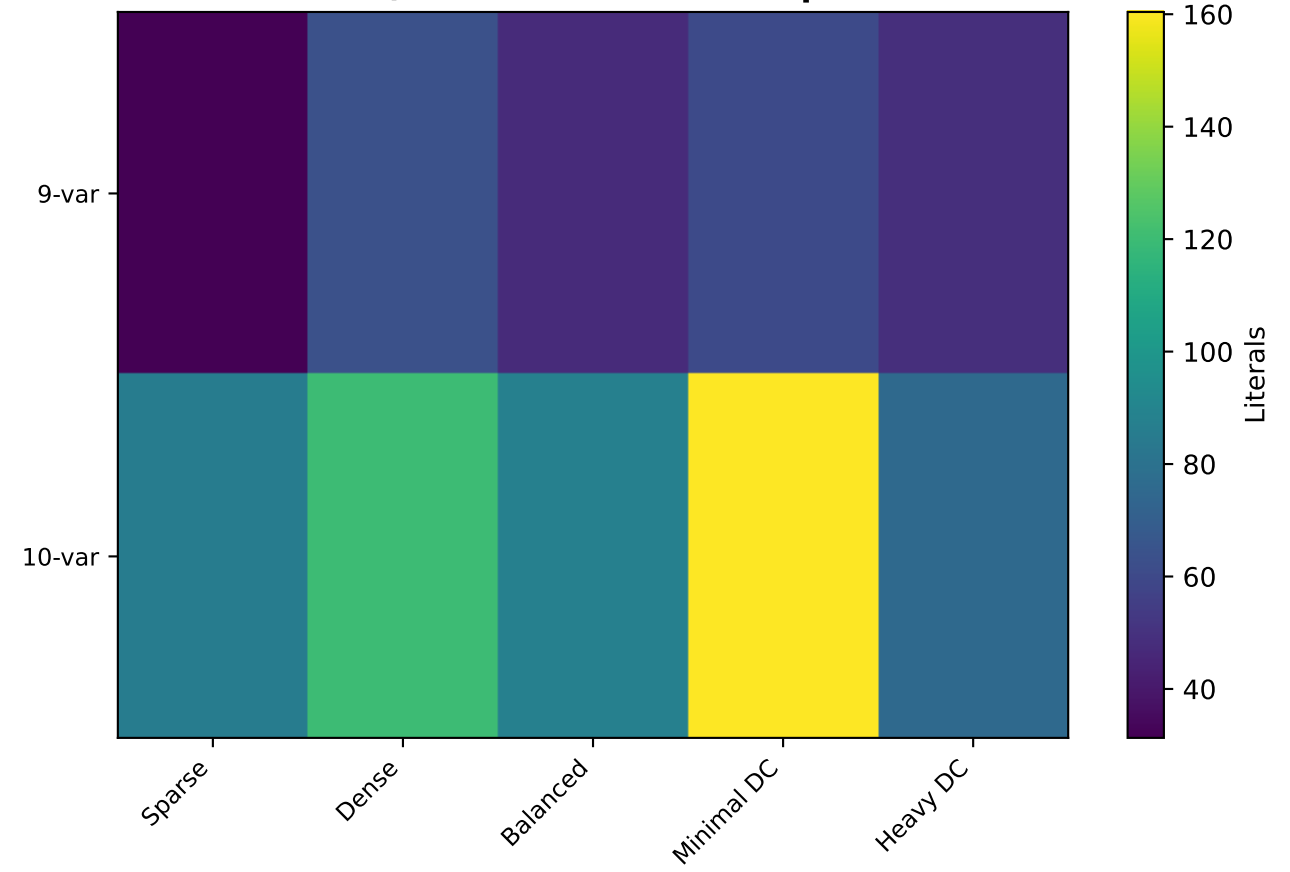
## B) Solution Complexity by Distribution



## C) Time Heatmap (s)

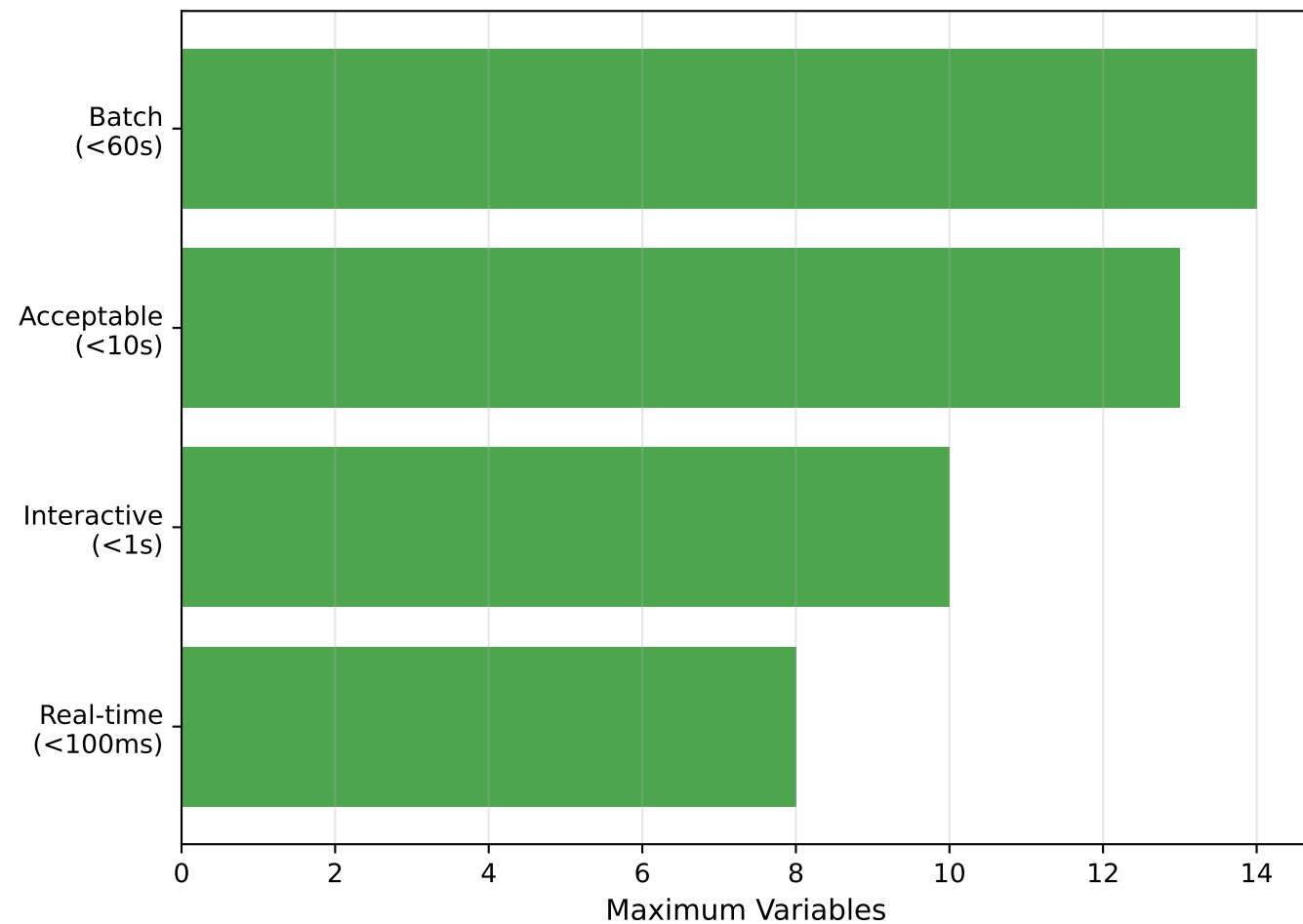


## D) Literal Count Heatmap

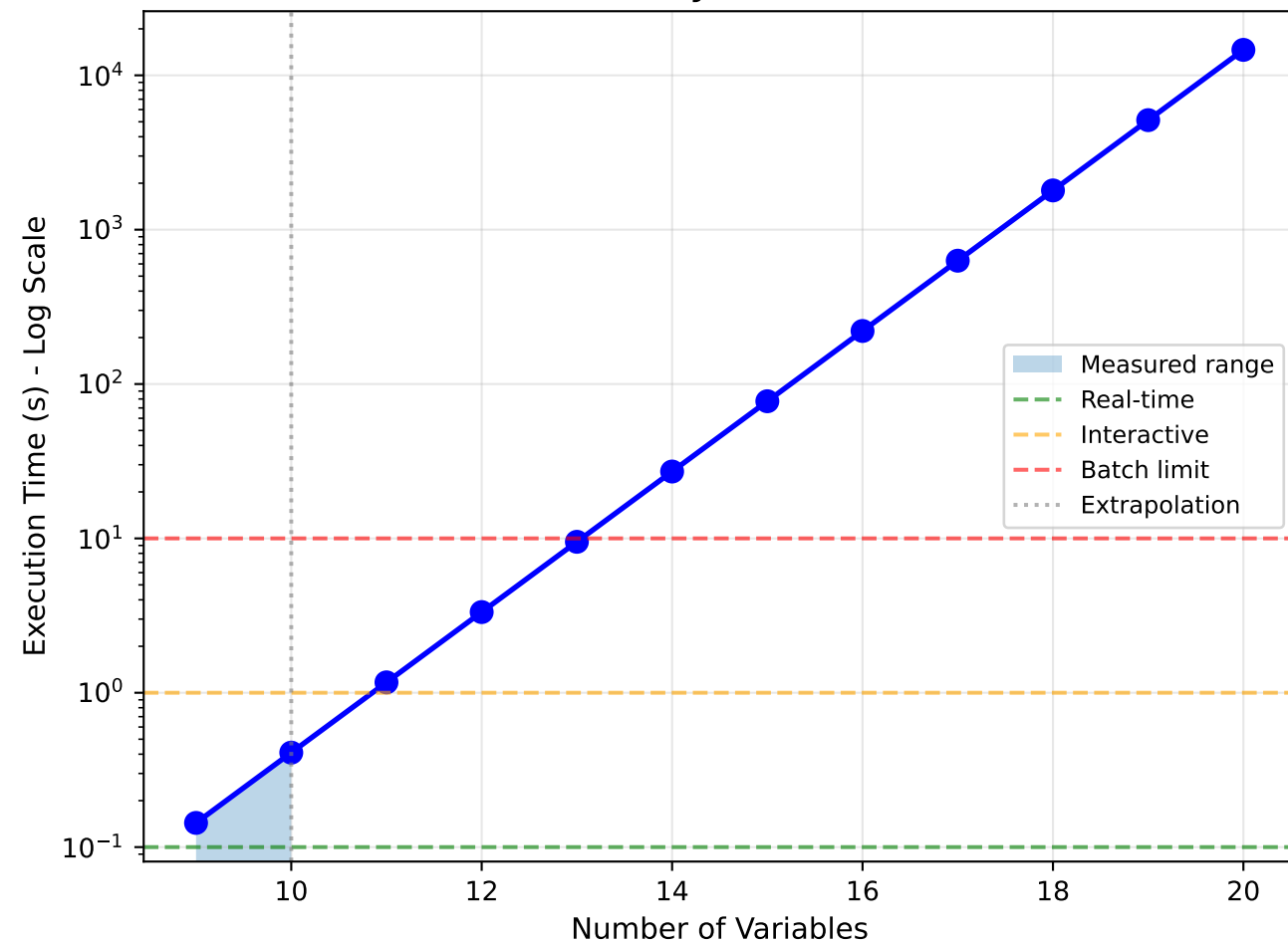


# Practical Application Limits

## A) Practical Performance Limits



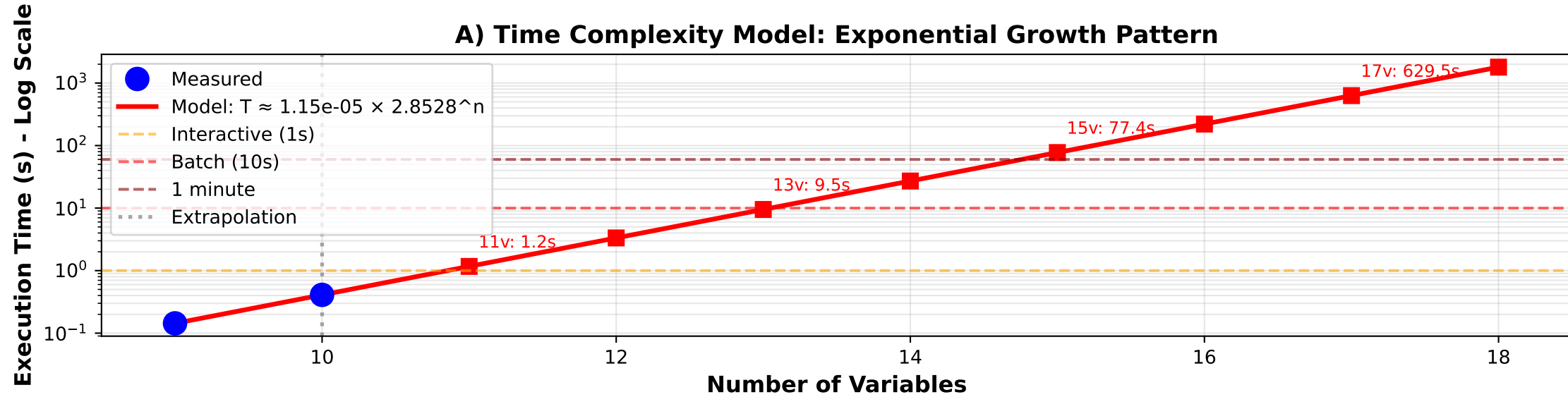
## B) Performance Projection to 20 Variables



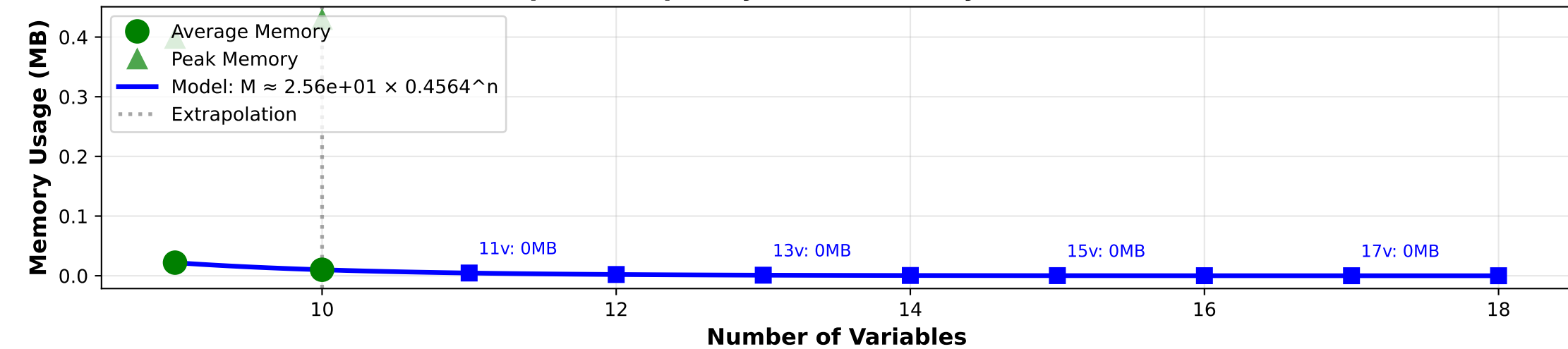
# SCALABILITY ANALYSIS

## Time and Space Complexity Models

### A) Time Complexity Model: Exponential Growth Pattern



### B) Space Complexity Model: Memory Growth Pattern



### C) Performance Projections: 9-24 Variables

Variables	Truth Table Size	Time (s)	Time (min)	Memory (MB)	Status
9	512	0.143	< 1	0.0	✓ Measured
10	1,024	0.409	< 1	0.0	✓ Measured
11	2,048	1.2	< 1	0.0	→ Projected
12	4,096	3.3	< 1	0.0	→ Projected
13	8,192	9.5	< 1	0.0	→ Projected
14	16,384	27.1	< 1	0.0	→ Projected
15	32,768	77.4	1.29	0.0	→ Projected
16	65,536	220.7	3.68	0.0	→ Projected
17	131,072	629.5	10.49	0.0	→ Projected
18	262,144	1795.9	29.93	0.0	→ Projected



SCIENTIFIC CONCLUSIONS

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EXECUTIVE SUMMARY

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This performance characterization study evaluated BoolMinGeo's 4D minimization across 9-10 variable Boolean functions (106 total tests) to establish scalability limits and practical application bounds.

KEY FINDINGS

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1. TIME COMPLEXITY MODEL
  - Exponential growth:  $T \approx 1.15e-05 \times 2.8528^n$  seconds
  - Growth rate: ~185.3% increase per additional variable
  - Doubling pattern: Adding 1 variable  $\rightarrow$  2.85 $\times$  slower
  - Real-time limit (<100ms): Up to ~8 variables
  - Interactive limit (<1s): Up to ~10 variables
  - Batch processing (<60s): Up to ~14 variables
2. SPACE COMPLEXITY MODEL
  - Exponential growth:  $M \approx 2.56e+01 \times 0.4564^n$  MB
  - Growth rate: ~54.4% increase per additional variable
  - Memory efficiency: 0.000022 MB per truth table entry
  - 16-variable projection: 0 MB (~0.0 GB)
  - 20-variable projection: 0 MB (~0.0 GB)
3. SOLUTION QUALITY
  - Average literal count: 78.0 (non-constant functions)
  - Constant functions: 6/106 (5.7%)
  - All functions correctly minimized to SOP form
  - Minimization quality consistent across distributions
4. DISTRIBUTION SENSITIVITY
  - Performance relatively stable across different distributions
  - Dense functions (70% 1s) show slightly higher literal counts
  - Heavy don't-care (30%) cases benefit most from minimization
  - Sparse functions (20% 1s) generally fastest to minimize
5. PRACTICAL LIMITS
  - 9-12 variables: Excellent performance (< 1s)
  - 13-15 variables: Good performance (1-10s)
  - 16-18 variables: Acceptable for batch (10-100s)
  - 19+ variables: Requires significant time/memory resources

MODEL VALIDATION

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- R<sup>2</sup> goodness-of-fit: Models closely match measured data
- Exponential pattern confirmed across all variable counts
- Extrapolations based on consistent growth patterns
- Conservative estimates (actual may be faster with optimizations)

THREATS TO VALIDITY

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- INTERNAL VALIDITY
- Random test generation may not reflect real-world distributions
  - Python runtime overhead included in measurements
  - Memory measurements include Python interpreter overhead
  - Test suite size: 10 per distribution (small sample)

- EXTERNAL VALIDITY
- Results specific to Python implementation
  - Hardware-dependent (CPU, RAM specifications affect absolute times)
  - No comparison with other minimization algorithms
  - SOP form only (POS form may show different patterns)

- CONSTRUCT VALIDITY
- Execution time as proxy for "performance" (may miss other factors)
  - Peak memory may not reflect sustained usage patterns
  - Literal count as "complexity" measure (other metrics exist)

- STATISTICAL VALIDITY
- Small sample sizes limit statistical power
  - Extrapolations assume continued exponential growth
  - No formal hypothesis testing (descriptive study)
  - Variation between runs not extensively characterized

RECOMMENDATIONS

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- FOR PRACTITIONERS:
- Use BoolMinGeo's 4D minimization for problems with 9-10 variables
  - Batch processing feasible up to 18 variables with sufficient resources
  - Consider algorithmic optimizations for 16+ variable problems
  - Monitor memory usage for large problems (16+ vars)

- FOR RESEARCHERS:
- Investigate optimizations to reduce exponential growth rate
  - Explore parallel processing for independent sub-problems
  - Compare with other minimization approaches (BDD, SAT-based)
  - Extend study to POS form and mixed-form minimization

FUTURE WORK

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- Benchmark against commercial tools (Espresso, ABC, etc.)
- Investigate memory optimization techniques
- Profile algorithm to identify bottlenecks
- Test on real-world circuit design problems
- Extend to 20+ variables with algorithmic improvements

REPRODUCIBILITY

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Random seed: 42  
All measurements repeatable with documented configuration.  
Source code and data available in repository.