

Evidence-Based Machine Learning Algorithm Selection for Construction Data Analytics: A Systematic Review

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Overview

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Research Problem

The Challenge:

ML applications in construction lack structured decision-making frameworks for algorithm selection

- Construction projects generate massive*, diverse datasets*
- Traditional analysis methods are insufficient
- No systematic approach to match algorithms with dataset characteristics
- Current practice: trial-and-error approach

Research Objectives

What We Investigated?

Primary Question:

Is evidence-based guidance being used to select ML algorithms in construction research?

We Analyzed:

- ✓ Frequency of ML algorithm usage
- ✓ Model objectives (prediction vs. classification)
- ✓ Reasoning behind algorithm selection
- ✓ Correlations with dataset characteristics

Methodology

Systematic Literature Review (PRISMA)

Data Sources:

- Web of Science
- IEEE Xplore
- ICONDA
- ScienceDirect
- ASCE

450

Initial Articles

70

After Screening

30

Final Articles

115

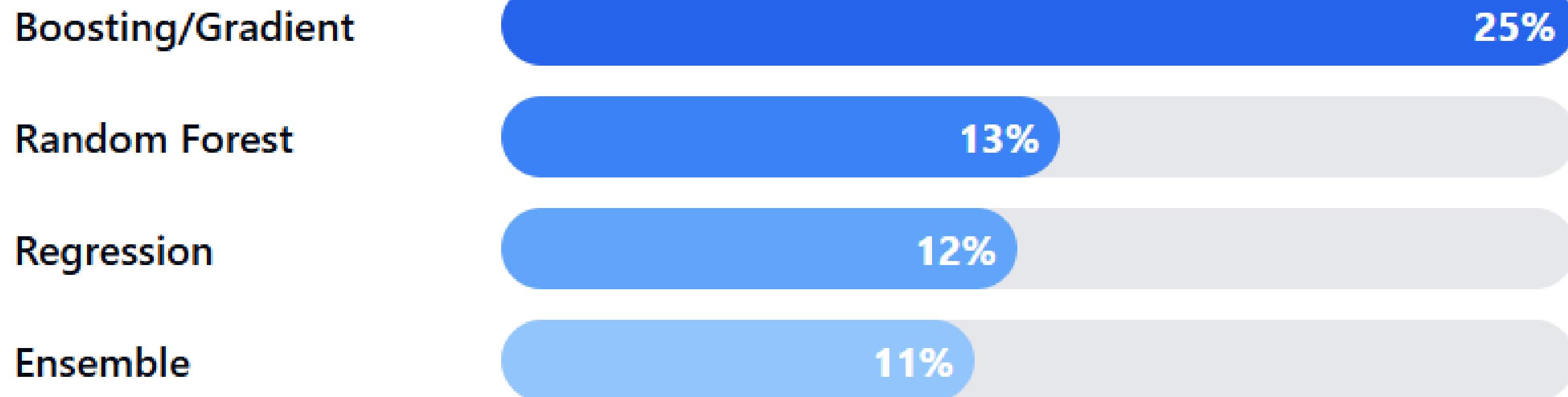
ML Methods

Search: "machine learning" AND "construction industry"

KEY FINDINGS

1. Algorithm Frequency

Most Frequently Used ML Algorithms



Key Insight:

Advanced ensemble methods dominate, suggesting construction datasets require sophisticated algorithms to handle complexity.

KEY FINDINGS

2. Algorithm by Analysis Objective

Prediction Models (55.7%)

1. Regression	73%
2. ANN	62%
3. Boosting/Gradient	54%

Classification Models (44.3%)

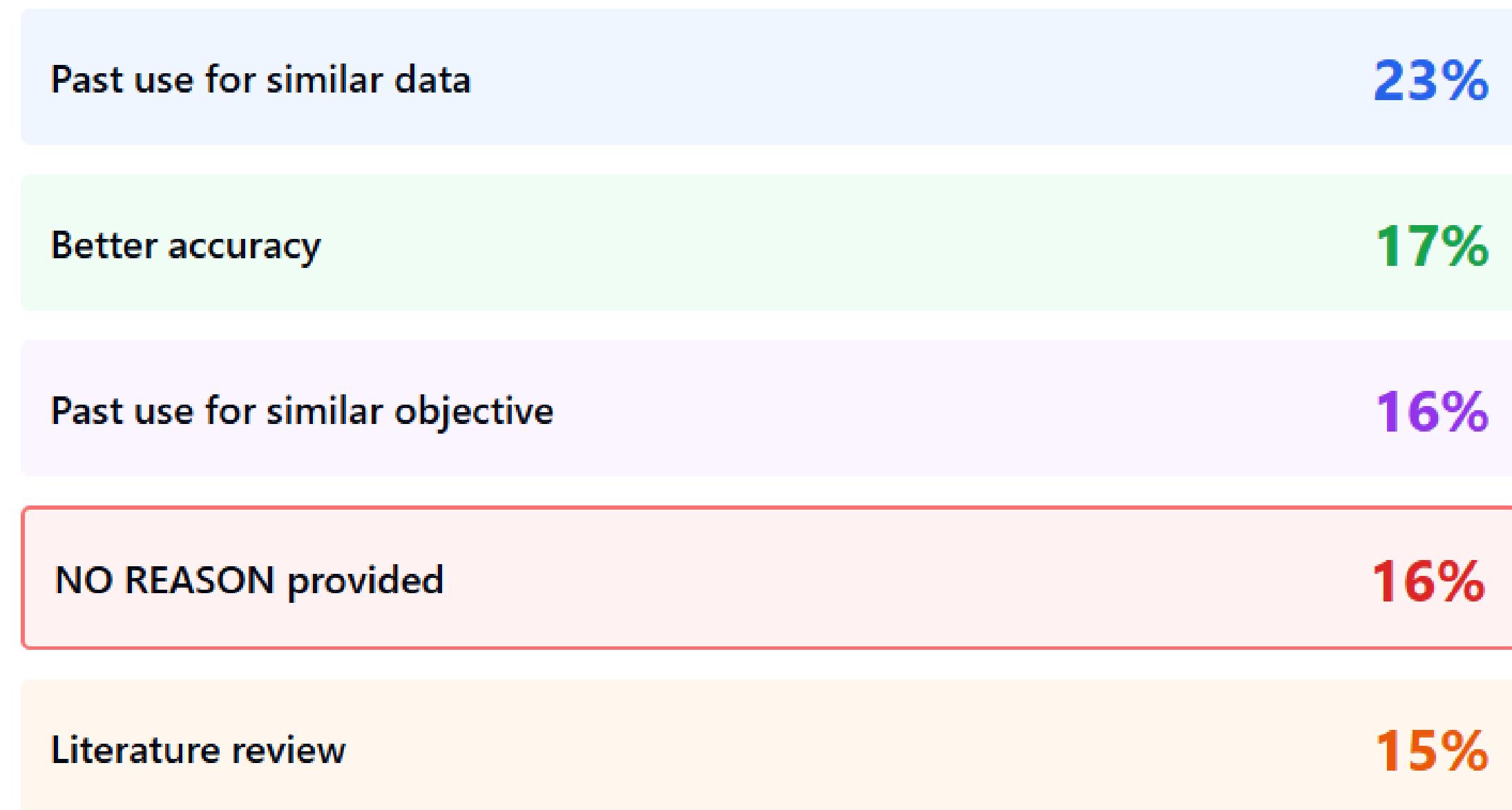
1. KNN	67%
2. Decision Tree	58%
3. SVM	56%

Note: Some algorithms like Random Forest and Ensemble show versatility across both objectives.

KEY FINDINGS

3. Selection Reasoning

Why Researchers Choose Algorithms



No systematic methodology!

CRITICAL FINDING:
39% rely on precedent alone
(past use) - not systematic
analysis
16% provide NO justification
whatsoever

The Problem Illustrated

Current Selection Approach

✗ Current Practice

- Trial-and-error testing
- Following precedent
- "It worked before"
- No systematic rationale
- Time-consuming
- May miss optimal solutions

✓ Needed Approach

- Evidence-based selection
- Data-driven decisions
- Match algorithm to data characteristics
- Consider analysis objectives
- Efficient selection process
- Optimal performance

The Problem Illustrated

3. Current Selection Approach

The Gap:

Researchers intuitively recognize algorithm-data relationships but lack a systematic framework to guide their decisions.

Algorithm Characteristics Summary

Boosting/Gradient

- + High accuracy, handles non-linear data, reduces overfitting
- Complex, computationally intensive

Random Forest

- + Robust, versatile, reduces overfitting
- Scalability issues with large datasets

Regression

- + Simple, interpretable, fast
- Assumes linear relationships, limited with complex data

ANN

- + Handles complex patterns, high accuracy
- "Black box", prone to overfitting

KNN

- + Simple, captures local patterns
- Poor with high-dimensional or imbalanced data

Decision Tree

- + Easy to interpret, handles mixed data types
- Prone to overfitting, sensitive to noise

Full details available in paper Table 1 and Section 4.1

Implication for Construction

Why This Matters?

For Researchers:

- ✓ Save time in algorithm selection
- ✓ Improve model performance
- ✓ Justify methodological choices
- ✓ Avoid trial-and-error approaches

For Industry Practitioners:

- ✓ Better prediction of costs, schedules, safety
- ✓ More reliable risk assessments
- ✓ Evidence-based decision making
- ✓ Efficient use of data analytics resources

For the Field:

- ✓ Standardization of ML practices
- ✓ Foundation for best practices
- ✓ Improved reproducibility

Recommendation for Future Research

What Next?

1. Selection Matrix Development

Create evidence-based norms mapping algorithms to dataset characteristics and objectives

2. Quadrant Framework

Categorize algorithms as traditional/regular/advanced based on complexity and timeline

Recommendation for Future Research

What Next?

3. Cross-Industry Analysis

Compare with retail, manufacturing, finance to identify algorithmic versatility

4. Empirical Validation

Test algorithm-dataset-objective combinations to validate optimal performance

Conclusions

Key Takeaways

Main Findings:

- ✓ Analyzed 115 ML methods from 30 construction studies
- ✓ Boosting/gradient methods most common (25%)
- ✓ Clear patterns by objective (Regression for prediction, KNN for classification)
- ✓ 39% rely on precedent, 16% provide no justification

Confirmed Gap:

No systematic selection approach exists that maps algorithms to dataset characteristics and analysis objectives

Thank you!

Questions?

Acknowledgements

- Name - Affiliation
- Name - Affiliation
- ETC.

Thank you

For any questions, please contact

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