

# Overview of Machine Learning Systems

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CHALLENGES AND CONSIDERATIONS IN APPLYING MACHINE LEARNING

STUDENT: PAULO EDUARDO DA SILVA JUNIOR



# Summary

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1. When to Use Machine Learning?
2. Machine Learning Use Cases.
3. Research vs. Production in ML.
4. ML Systems vs. Traditional Software.
5. Conclusion.

# When to Use Machine Learning?

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1. Identifying complex, predictable patterns in data.
2. Advantage in predictive tasks such as recommendations and fraud detection.
3. Assessing need, ethics, and cost-benefit.
4. Machine learning is an approach to
  - Learn;
  - Complex patterns from;
  - Existing data and use these patterns to make;
  - Predictions on;
  - Unseen data.

# Machine Learning Use Cases

Industries:

- Enterprise: Customer insights, process automation, anomaly detection.
- End-user: Personal assistants, machine translation, image recognition.

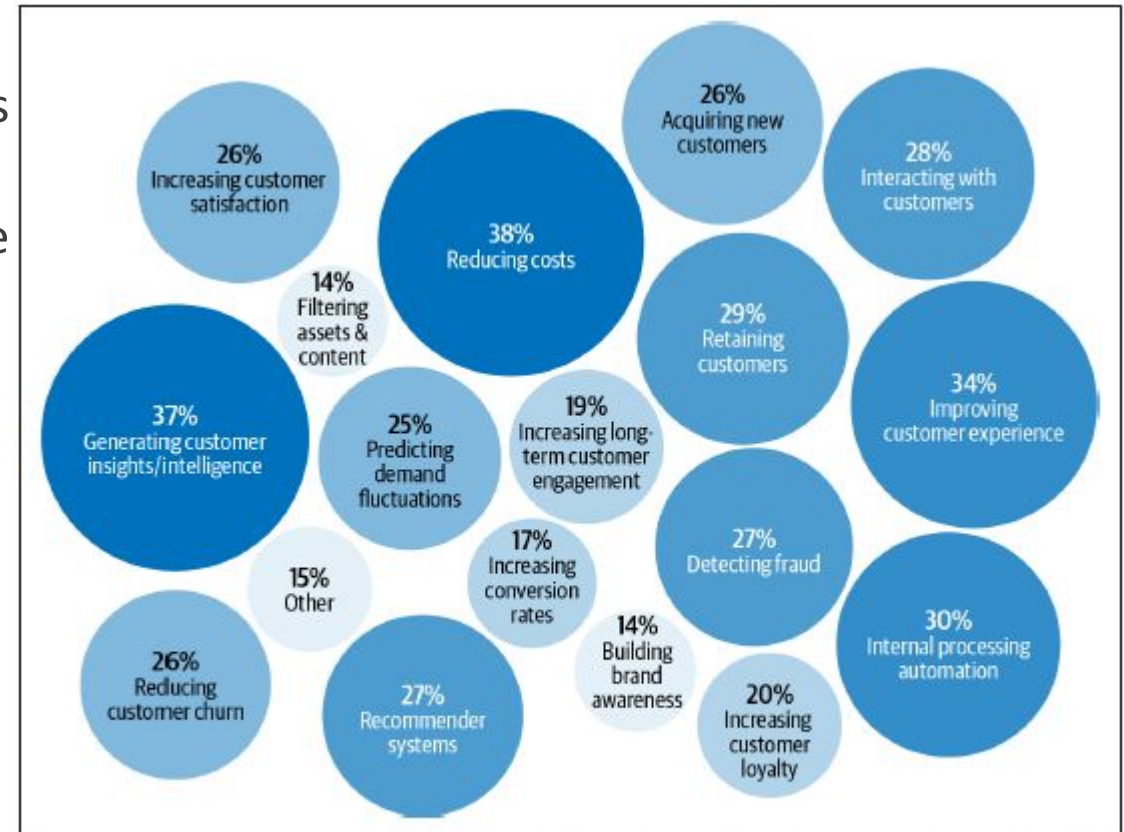


Figure 1-3. 2020 state of enterprise machine learning. Source: Adapted from an image by Algorithmia

# Research vs. Production in ML

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Points:

- Challenges in dynamic data and latency.
- Need for interpretability and fairness.
- Stakeholder demands: engineering, product management, and sales.

*Table 1-1. Key differences between ML in research and ML in production*

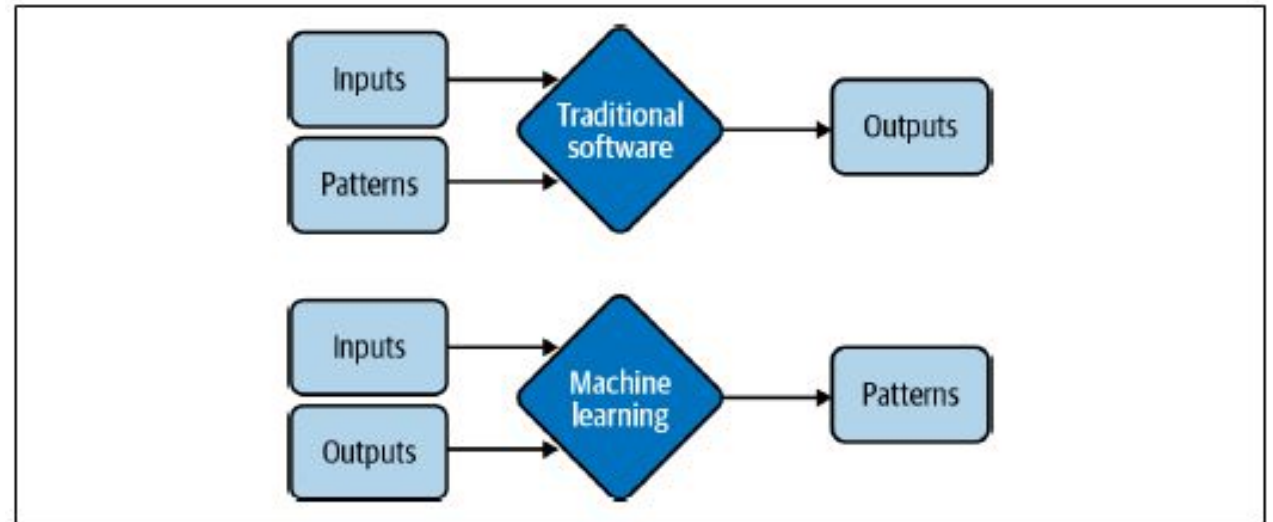
	Research	Production
Requirements	State-of-the-art model performance on benchmark datasets	Different stakeholders have different requirements
Computational priority	Fast training, high throughput	Fast inference, low latency
Data	Static <sup>a</sup>	Constantly shifting
Fairness	Often not a focus	Must be considered
Interpretability	Often not a focus	Must be considered

# ML Systems vs. Traditional Software

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Comparison:

- ML: Dependence on data and models.
- Traditional Software: Main dependence on code.
- Challenges of adaptation, monitoring and maintenance in ML.



*Figure 1-2. Instead of requiring hand-specified patterns to calculate outputs, ML solutions learn patterns from inputs and outputs*

# Conclusion

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## Summary:

- The complexity of ML systems demands more than algorithms.
- Infrastructure, monitoring, and maintenance are essential.
- Future chapters will detail components for a complete ML system.