Projects in Artificial Intelligence: Software 2.0

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Team



Planning



Data



Tooling and Infrastructure



Deployment

"87% of data science projects never make it into production"

- Venture Beat

Why?

- Technically infeasible
- Lack of Collaboration
- Unclear Success Criteria
- "The Value of the Project Does Not Outweigh its Technical Complexity" (Tobin, 2022)

Machine Learning: The High-Interest Credit Card of Technical Debt (Sculley, et al)

- Complex Models Erodes boundaries
- Data Dependencies Cost More than Code Dependencies
- Feedback Loops
- Anti-patterns
- Configurational Debt
- Subject to instability



Can we identify high impact, low-cost problems that can benefit enough to outweigh the complexity cost that ML brings to development?



Cost drivers Problem difficulty Accuracy requirement Data availability

Main considerations

- · Is the problem well-defined?
- Good published work on similar problems? (newer problems mean more risk & more technical effort)
- · Compute requirements?
- · Can a human do it?
- How costly are wrong predictions?
- How frequently does the system need to be right to be useful?
- · Ethical implications?
- How hard is it to acquire data?
- How expensive is data labeling?
- · How much data will be needed?
- How stable is the data?
- Data security requirements?

The Rise of Software 2.0: Data Driven Development

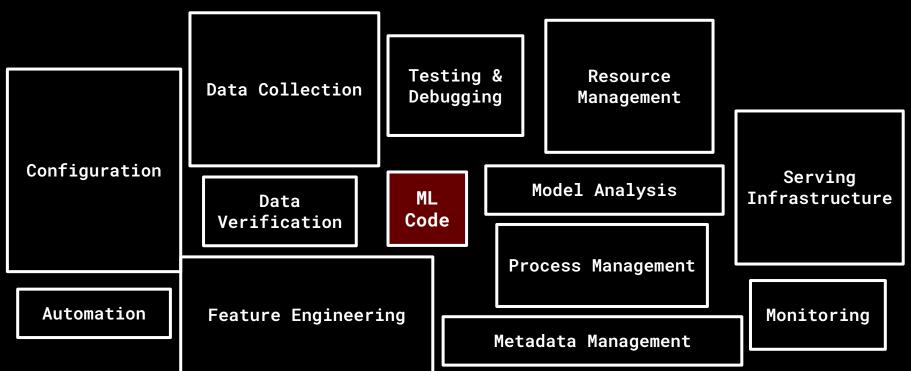
2.0 Developers (Data)

- Curation
- Updates & Growth
- Transforming
- Cleaning

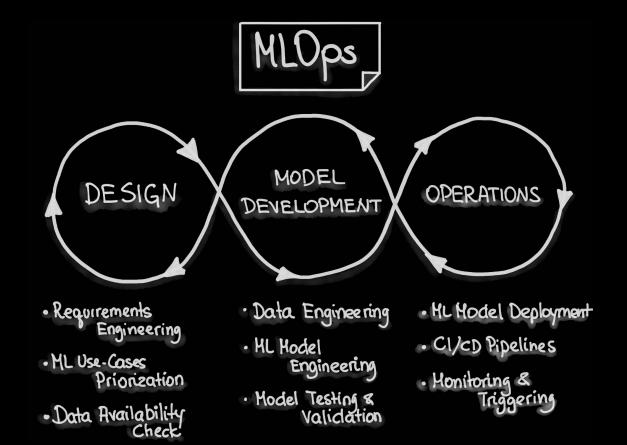
1.0 Developers (Traditional)

- Training Infrastructure Development
- Analytics
- Visualization
- Labeling















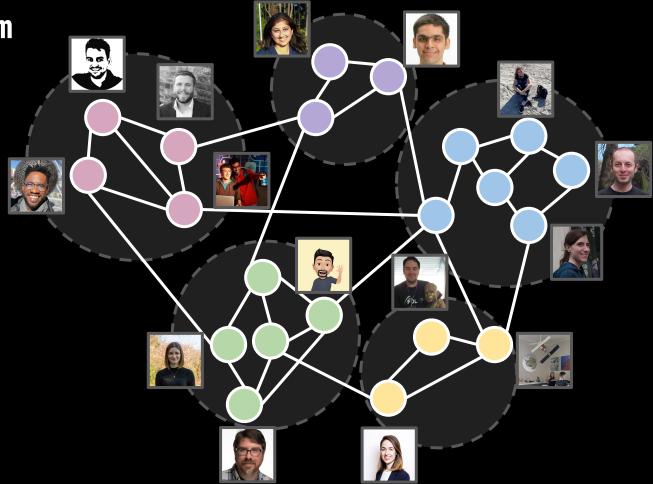








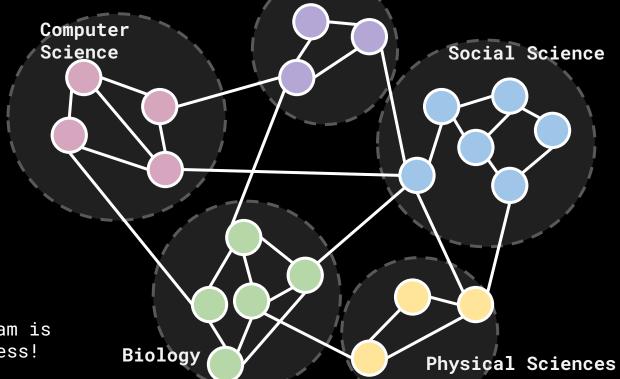
Team





Team

Engineering



Diversity of Team is the key to success!



- Diversity of Background
- Openness to learning
- Strong Subject Matter Expertise
- Strong Computer Science Skills
- Collaboration



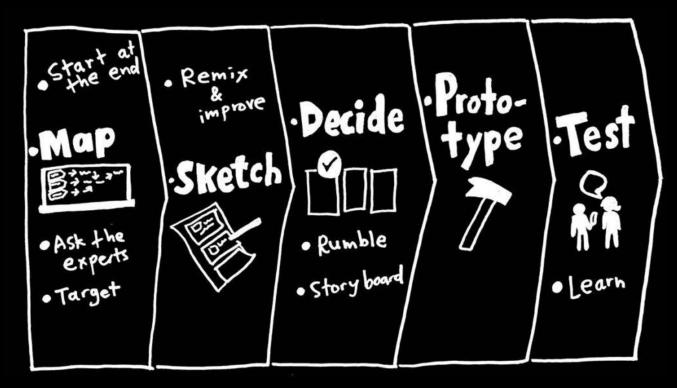
From Prototype to Production

- Problem Refinement
- Collect data
- Clean and preprocess data
- Explore the data (EDA)
- Feature and model selection
- Train model
- Evaluate results
- Deploy



How might we use Agile techniques to plan our project?









François Chollet @ @fchollet · Jan 24

ML researchers work with fixed benchmark datasets, and spend all of their time searching over the knobs they do control: architecture & optimization. In applied ML, you're likely to spend most of your time on data collection and annotation — where your investment will pay off.

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23

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Machine Learning Needs Data.



- What prediction problem would you like to solve?
 - Classification
 - Regression
- What kind of data is needed to solve the problem?
 - Time series data
 - o Image data
 - ⊤ext data
 - Sound
 - Tabular



Open Source Datasets

- Kaggle
- UCI Machine Learning Repository
- US Government Data
- Open Data on AWS
- Google Cloud Public Datasets
- Or create your own...



Tooling & Infrastructure

"All-in-one"

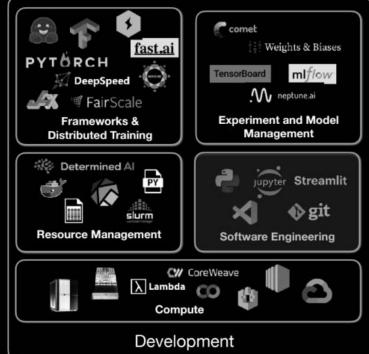














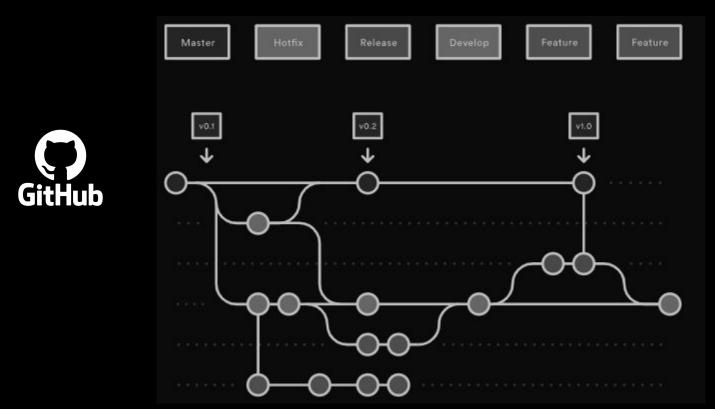
Tooling & Infrastructure

Continuous Integration and Development

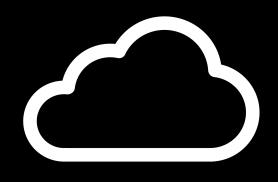
- Data versioning
 - landing (raw)
 - staging (ML ready)
- Model versioning
- Results/Feature Deployment



Tooling & Infrastructure









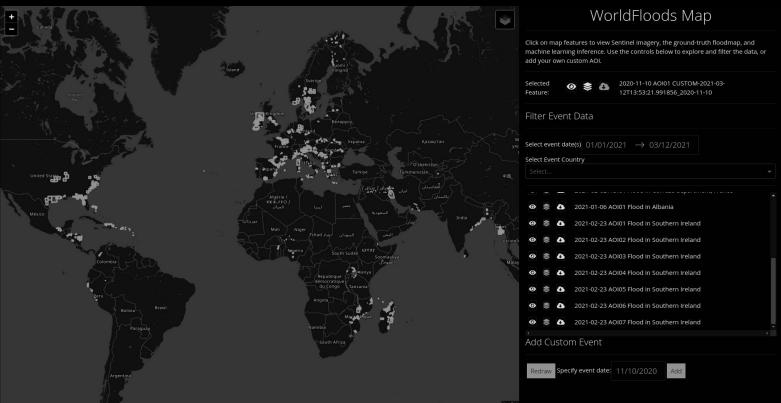


Cloud

Payload

Mobile





Thank You!









