

Agenda

- About Me and Springboard
- Current State of the AI Talent Market
- Common Approaches to Solving the Problem
- Innovative Models to Consider



About Me



Gautam Tambay
Co-founder and CEO, Springboard

@gautambay @springboard





Experts in Learning with a Global Footprint

+10,000 Alumni and Mentors at the World's Leading Companies:

VISA IM SAMSUNG Dropbox amazon.com

Deloitte. accenture verizon

UBER McKinsey airbnb Company

facebook ORACLE UA AdRoll

one cisco pandora zomato



The training paid insane dividends and helped us become thought leaders within our company and industry. Things we built during the course are part of our process now and we're winning business because of it.



Kevin TroyanosSVP Marketing Analytics
Saatchi & Saatchi

A Common Refrain

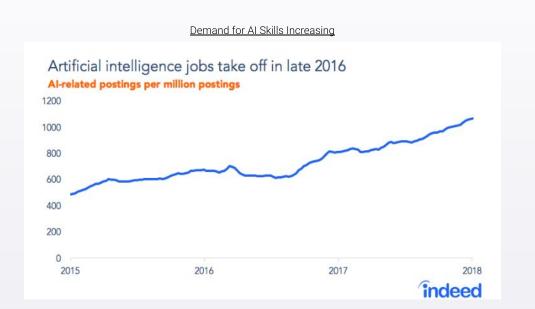
- Conly 10% of our engineers are truly prepared to take advantage of advanced ML and Al techniques to support the business. It limits how we can use these techniques to solve critical problems
 - Al Engineering Manager Global Logistics Company

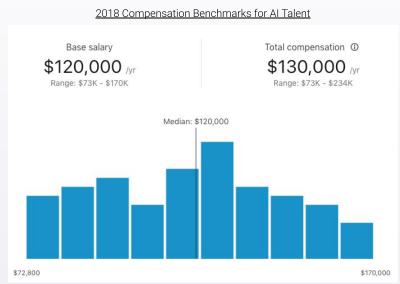
- engineering roles are the hardest engineering roles for us to fill. I don't see that changing for a while. Demand is super high. Everyone wants and needs to hire these people"
 - ML Program Lead
 Software Company

- Our executives don't understand how to interpret and make decisions based on the Al models. It's an unfamiliar way of thinking for many of them."
 - Managing Director

 Top-tier Consulting Firm

Job Growth and Salaries Continue to Rise







Common Solutions, Common Results

- Increase compensation
- Lower the quality bar
- Hire third-party recruiters
- Hackathons, conferences, and university partnerships
- Stretch current resources



Two Innovative Approaches



Tech College

Uber

MLaaS Platform

Case Study: Capital One Bank

Challenges faced:

- Demand for hiring ML talent outpaced supply
- Market competition for talent was reaching unsustainable levels

Businesses Impacted:

- Core Business: ML enhancing the banking experience.
- Fraud detection and mitigation
- New Products: Eno App; chat bots





Introducing Tech College





Overview

- Launched in 2017
- First pilot: part-time ML training for 350 engineers
- Long-term ambition: offer technology courses for all +50,000 employees

Approach to Learning

- Built by engineers, for engineers
- Breadth and depth
- Blending learning
- Learning from our leaders
- Learning for all

Banking On It



Being known as a company that supports and encourages employees in pursuing lifelong learning and professional development is a big part of what helps us attract such talented people to work for us.

> - George Brady Chief Technology Officer





Outcomes

- Campus recruiting improved in both quality and quantity of applicants
- Attrition of HiPo engineers declined as compelling career paths emerged.
- Better results: 60% decrease in model development times
- Consistently ranked as one of the best places to work among U.S. companies.





Case Study: Uber

Uber



Challenges faced:

- No system to build reliable and reproducible pipelines for managing training and prediction data at scale.
- Teams building bespoke one-off solutions for using models in production
- Training models were limited in size

Businesses Impacted:

- Accurate marketplace forecasts
- Uber Eats (Recommendations, prep time, ETAs)
- Predictive one-click chat
- Faster customer support
- Driver ETAs and ride safety check
- Self-driving cars



Michelangelo: Uber's Internal MLaaS Platform

Core Strategic Pillars of Michelangelo ML Platform



[Michelangelo] is a ML-as-a-service platform that democratizes machine learning and makes scaling AI to meet the needs of the business as easy as requesting a ride.

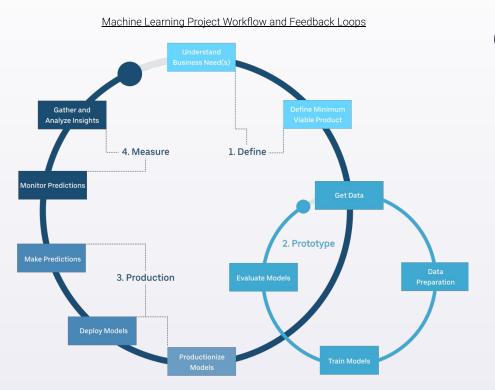
It enables internal teams to seamlessly build, deploy, and operate machine learning solutions at scale. It is designed to cover the end-to-end MI workflow

- Jeremy Hermann and Mike Del Blaso Machine Learning Platform Team





Michelangelo: Uber's Internal MLaaS Platform



A high priority for the Michelangelo team is enabling data science teams to go faster. The faster we go, the more experiments we can run, the more hypotheses we can test, the better results we can get. We are constantly thinking about this process and tightening the feedback loops so it's easier and faster to do iterative and agile data science.

- Jeremy Hermann and Mike Del Blaso Machine Learning Platform Team



Outcomes

- At any given time, hundreds of use cases representing thousands of models deployed in production
- Millions of predictions made every second
- Increased adoption of ML techniques
- Attracts great talent who want to work on ML/Al projects

Uber





Questions?

Send us an email at hello@springboard.com