

# INTELLIGENT SOCIAL RECOMMENDATIONS

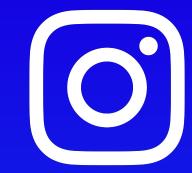
Thomas Bredillet (tnb@fb.com)

# AGENDA

What are we trying to solve

2 What we learned

3 What's next?



# WHAT ARE WE TRYING TO SOLVE

#### PROBLEMS

How can we use intelligent algorithms to improve the Instagram experience? What are they key issues where A.I can help our users?

# Never miss friends and family content

Reverse chronological ranking means you'll see content from whomever posts most frequently

#### Surface quality

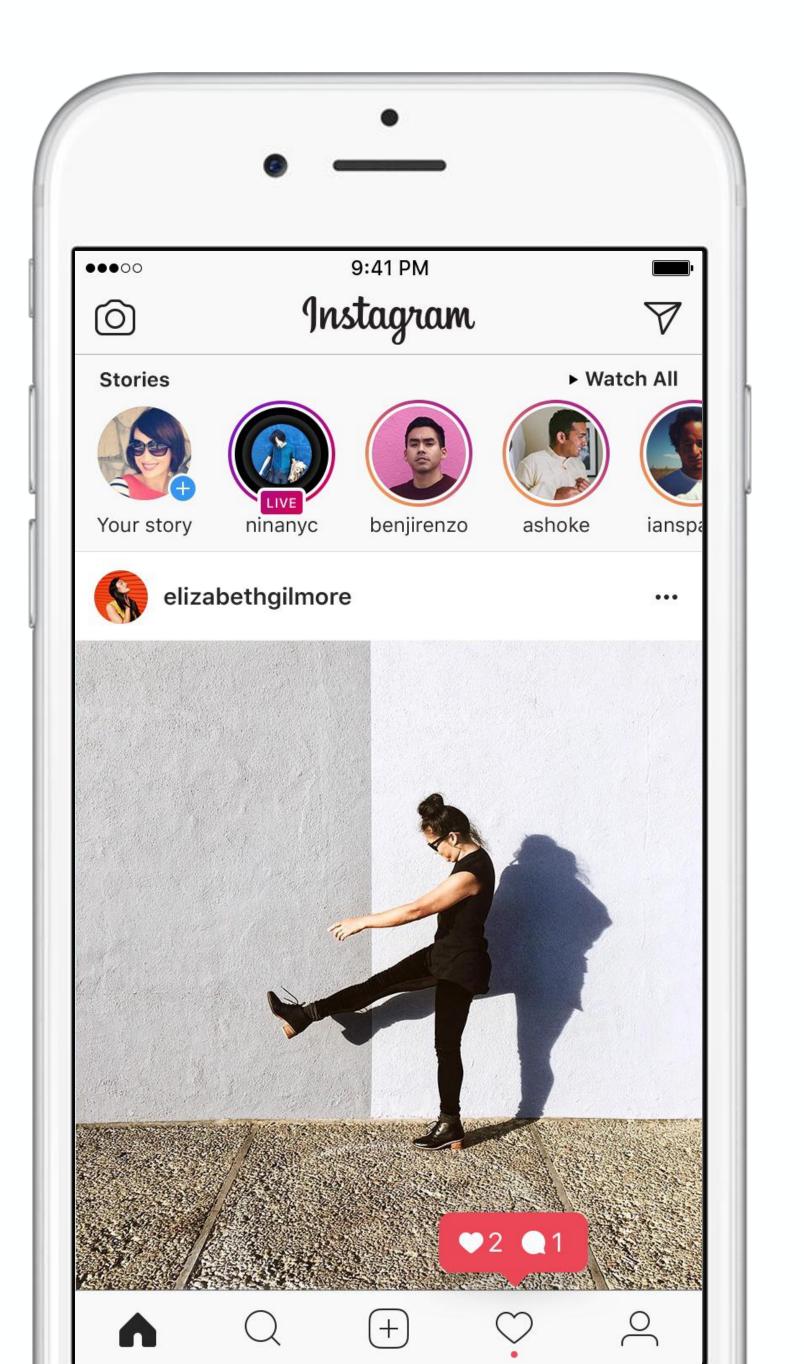
Amongst all the new available content we need to decide what stands out for you personally

#### Source hashtags

Following a hashtag means we have to find the best K new medias amongst an almost limitless pool of great content

#### Understanding the context

Could we improve our users' experience by factoring things such as cellular reception?



#### DOES FEED RANKING WORK?

Yes!

#### Engagement

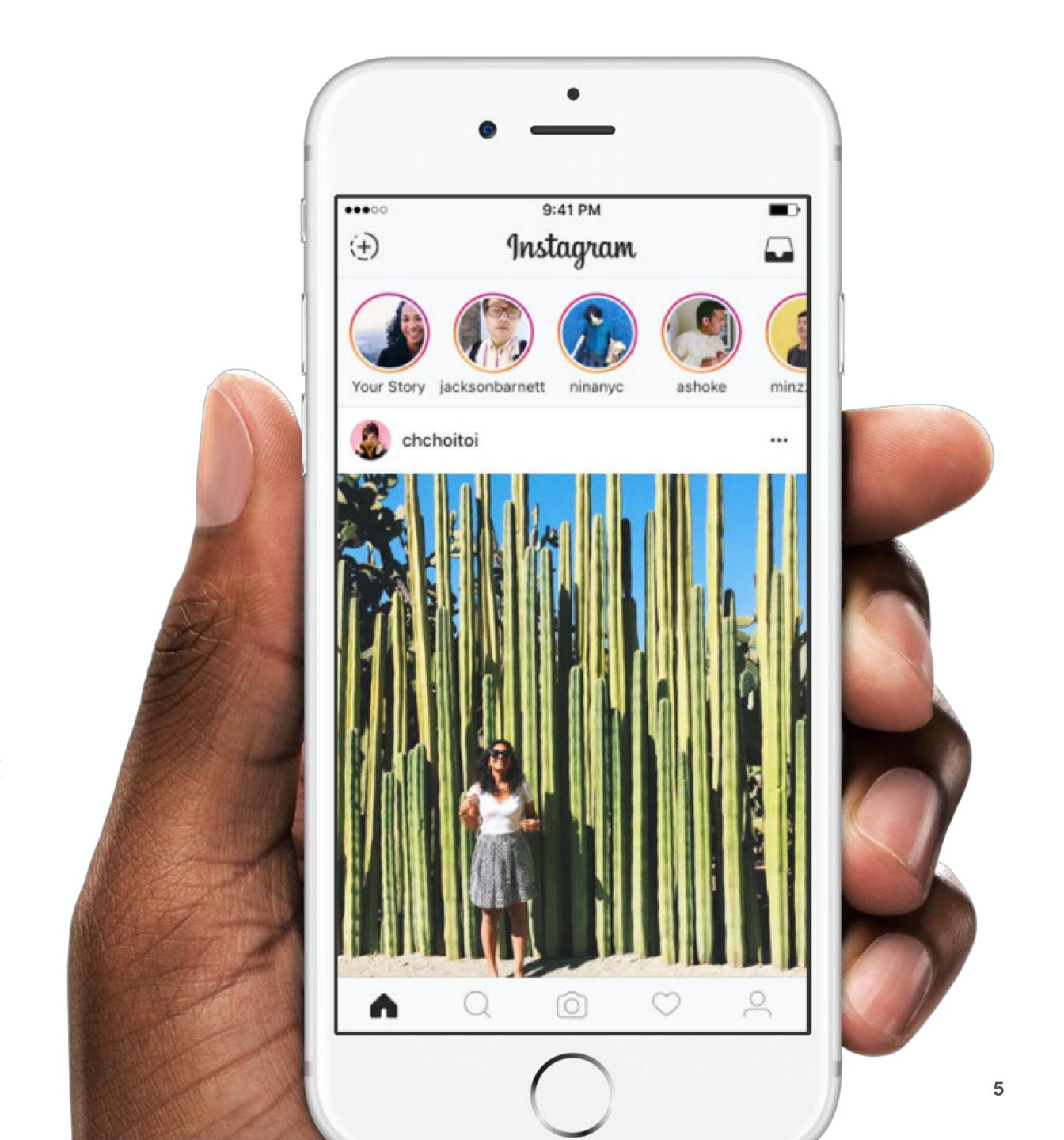
Getting an A.I ranking feed means you will engage with significantly more content than if you get chronological ranking.

#### Content

People getting our algorithms see more posts from their friends and family.

#### Sentiment

A strong sign that our machine learned models are providing a much better experience is that users sentiment is much higher with our intelligent feed

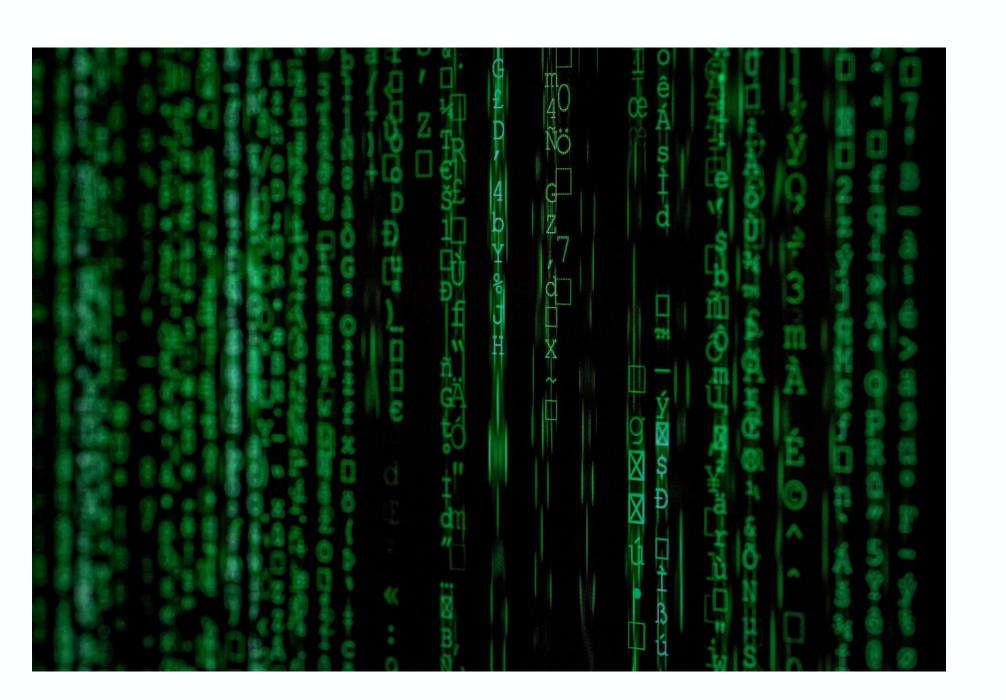




# WHAT WE LEARNED

#### DATA

- It's all about setting up the proper problem to enable machines to learn and generalize
- Data storage and organization at scale is hard
- Effective client logging is key, those are the atomic components and the ground truth for our algorithms
- Content understanding helps relevancy



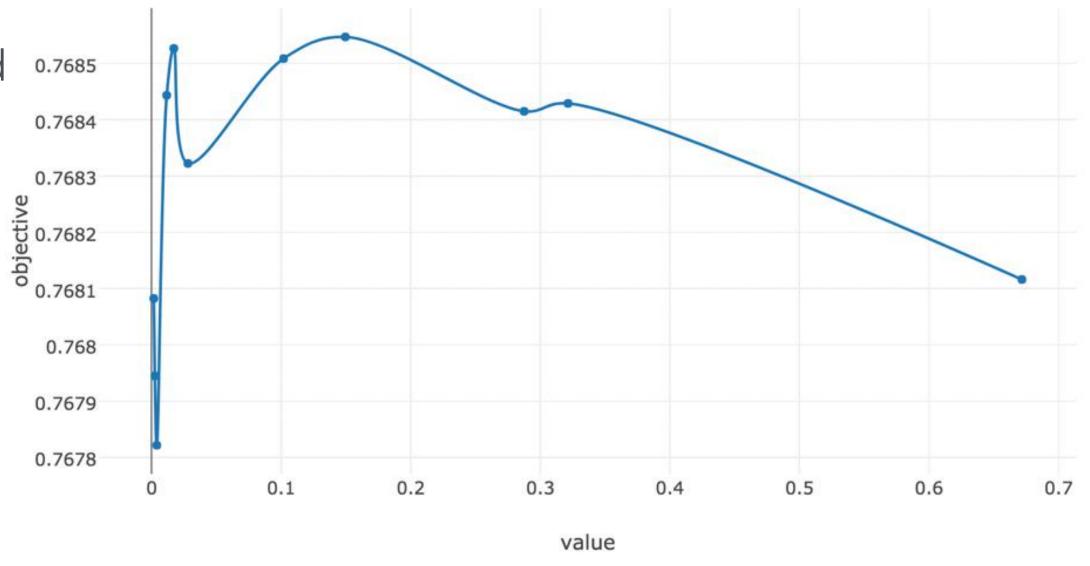
### DATA FRESHNESS AND TRENDS

The importance of stationarity!

 Helping our algorithms learn properly over time as the world evolves

Avoiding A.I feedback loops

inputs.training\_config.algorithm\_config.lambdamart.boosting\_config.learning\_rate



## EXPERIMENTATION - SMALL EFFECTS

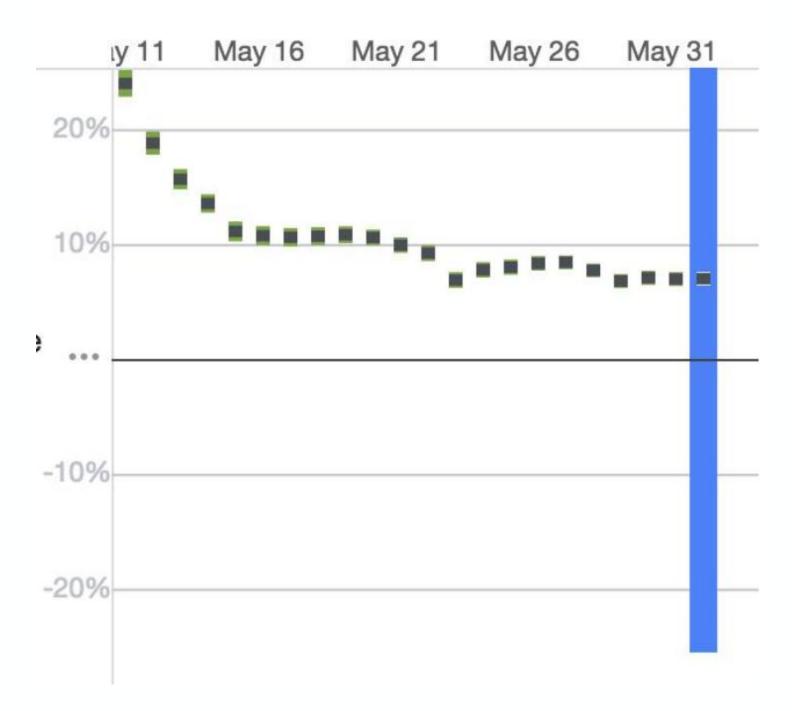
- What represents an experiment in the world of A.I?
- Solid best practices and scientific method
- Trade off between iteration speed and accuracy
- A.I informs business learnings and ecosystem understanding!
- Scaling ML teams



Photo by Louis Reed

## NOVELTY EFFECTS

- Novelty effects are hard to factor in our A.I pipelines
- Some experiments are more prone to novelty effects than others
- We need proper experiment termination guidelines



#### DIFFERENT SHAPES FOR A.I

- A solid heuristic shouldn't be discounted too fast!
- Trade-offs between sophisticated Machine Learning and light weight statistical modeling
- Holdouts are key to any A.I enabled systems
- Diminishing returns vs research efforts

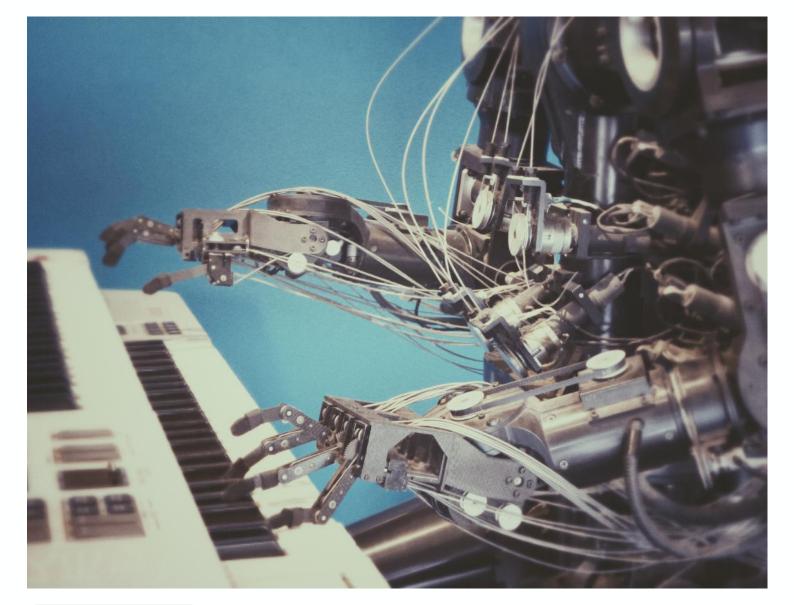
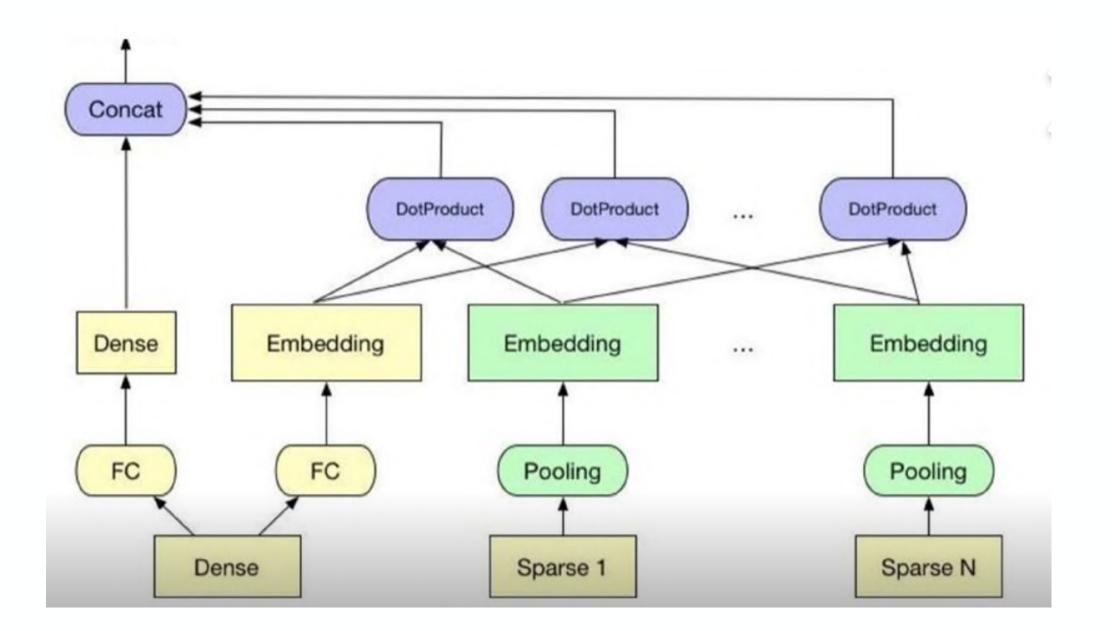


Photo by Franck V.

## POINT WISE MODELS

- Understanding our users
- Feedback is a gift!
- Optimizing for the right business goals



## VALUE MODELING

#### Where the business logic goes

- We need to combine our different probabilities in one final score
- The value model is where human intervention meets A.I.
- Enforced business logic

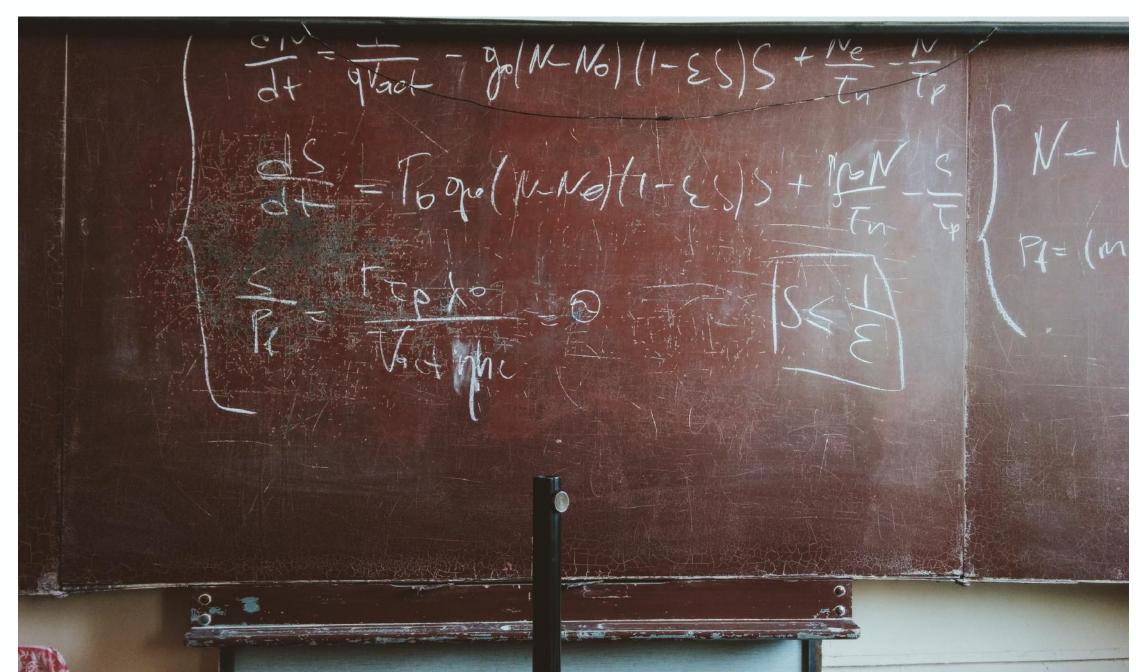
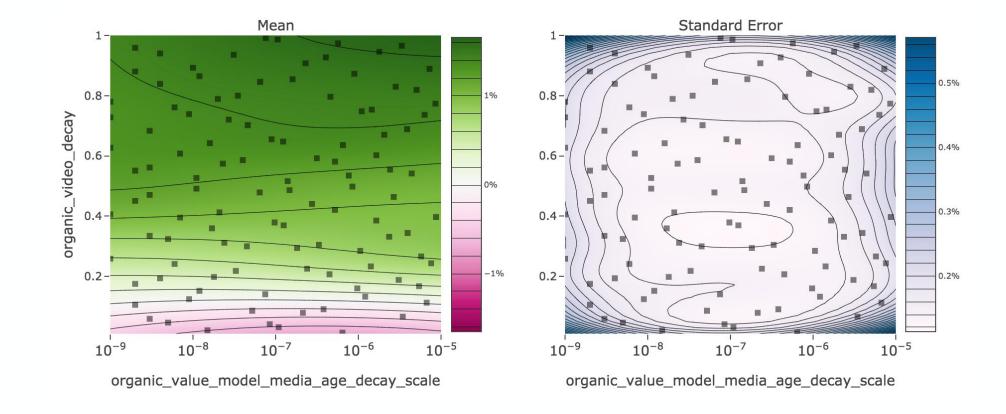
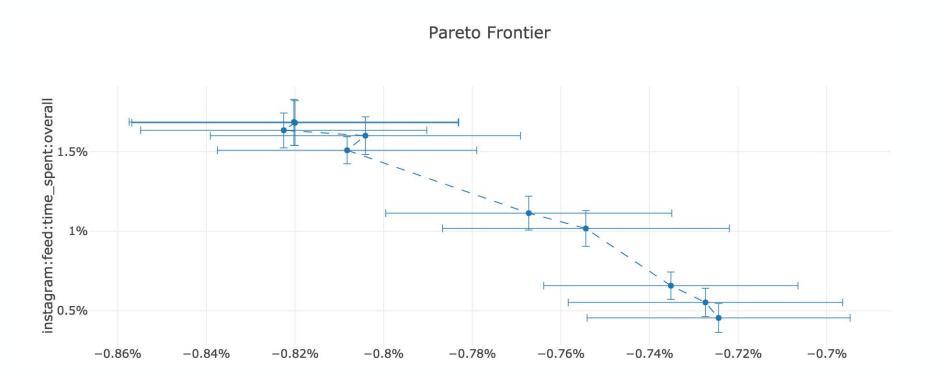


Photo by Roman Mager

## BAYESIAN HYPER PARAMETER TUNING

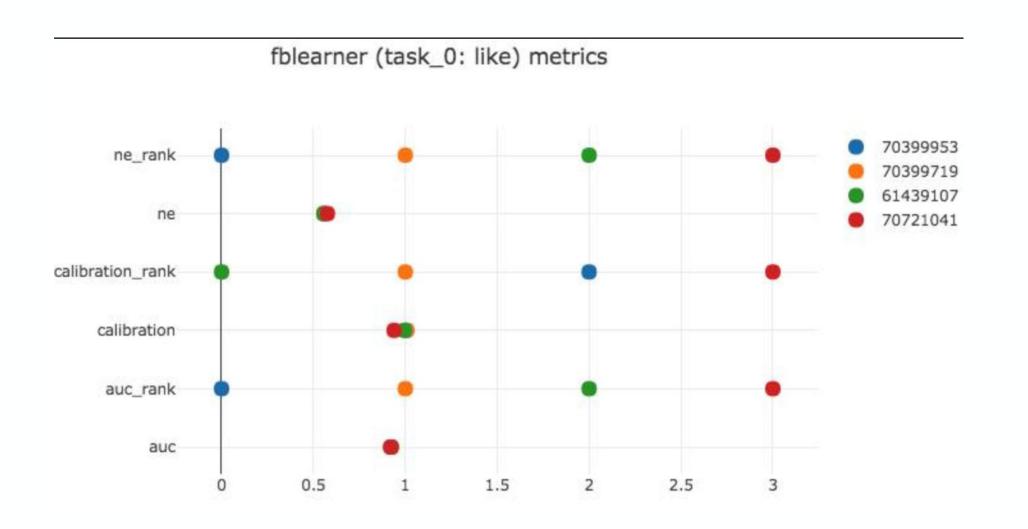
- Having an A.I to overview other A.Is
- Processing and constantly tuning our signals
- Giving us clear business trade-offs between metrics we care about
- Providing us levers and knobs to incentivize specific behaviors





## ITERATION SPEED - OFFLINE ANALYSIS

- Backtesting framework understanding ecosystem effects
- Deep diving into deep black boxes
- Measuring effectiveness of our algorithms offline without disruption of our users' experience





# WHAT'S NEXT?

# LOOKING AHEAD

- Personalization
- Incorporate content signals
- Memory effects

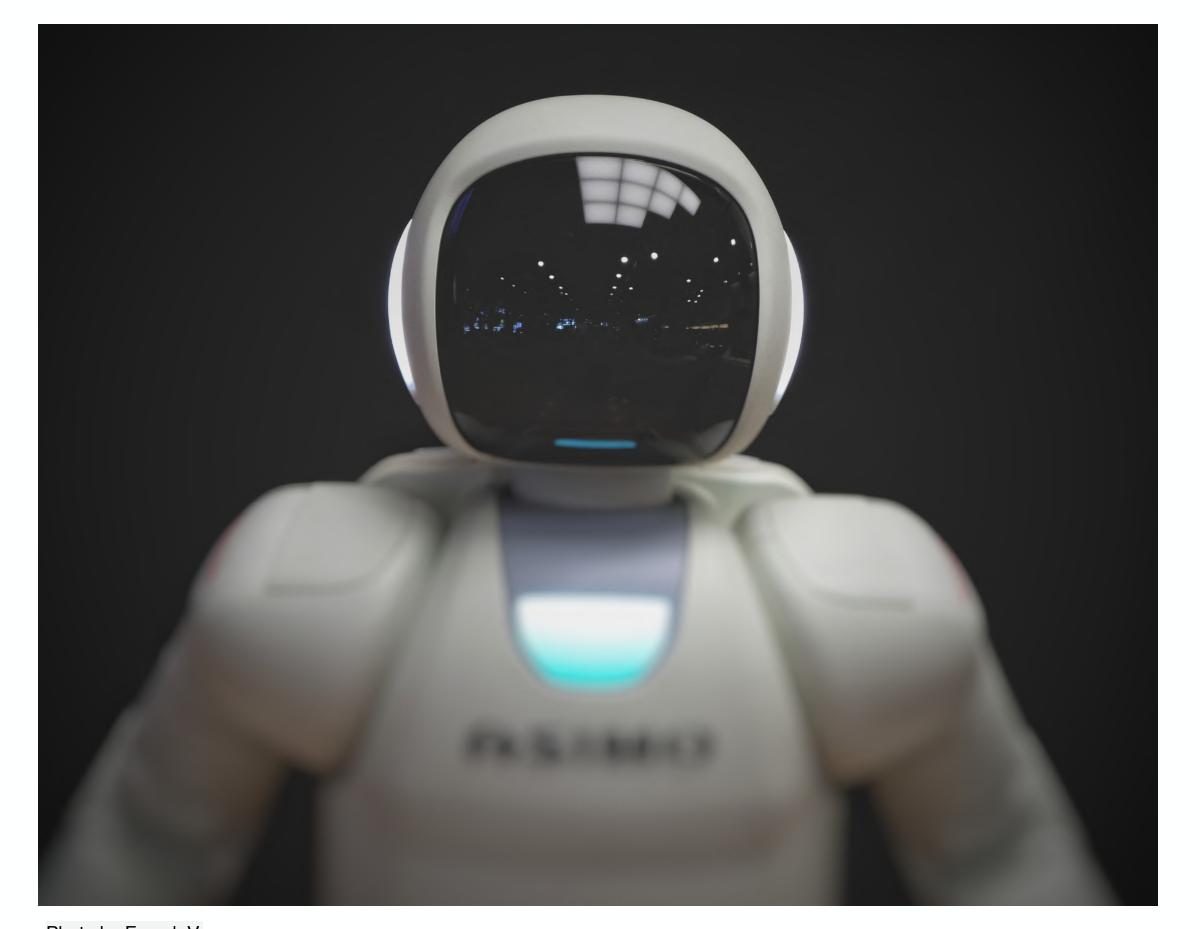


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