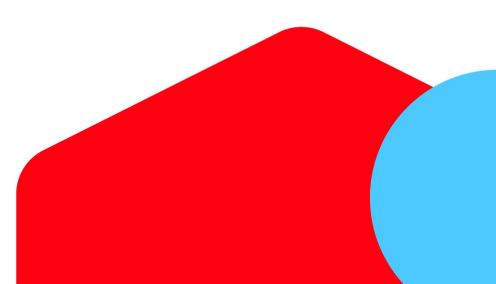
Building MLOps Infrastructure at Japan's Largest C2C E-Commerce Site

Platform, productionalization, and monitoring

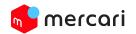
Teo Narboneta Zosa Ryan Ginstrom Search Team

mercari

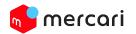


Contents

- 1. Introduction
- 2. The Problem
- 3. System Evolution
- 4. Future Directions
- 5. Conclusions



About Search at Mercari



About Mercari

- Japan's largest consumer-to-consumer (C2C) online marketplace
- FY2022 numbers:*
 - Gross merchandise value (GMV): ~¥880 billion (~\$6.7 billion USD)
 - Net sales: ~¥150 billion (~\$1.1 billion USD)





Search at Mercari

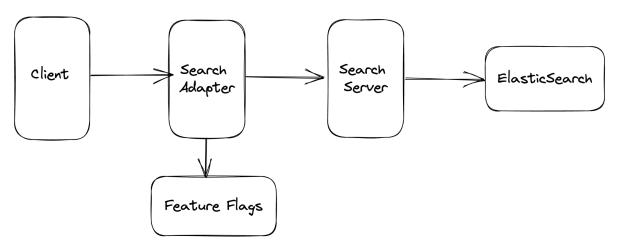
- Over 20 million monthly active users (MAU)
- 100s of millions of active listings in catalog
- 1,000s of queries per second (QPS)





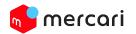
Search Topology

- "Traditional" term-based search
 - Lucene/Elasticsearch





The Problem



Problem

Blind spots (that AI can see)

- Ambiguous keywords
- Semantics ("cool toys for boys")
- Personalization

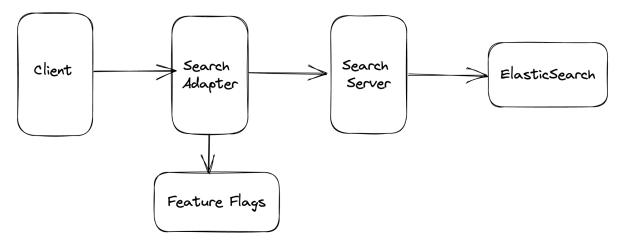




The Problem

Integrating ML into a "traditional" term-based search architecture

- Classic search infrastructure and workflow; no "easy hooks" for AI
- Latency budget: 10's of ms



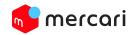


The Problem

Integrating ML into a "traditional" term-based search architecture

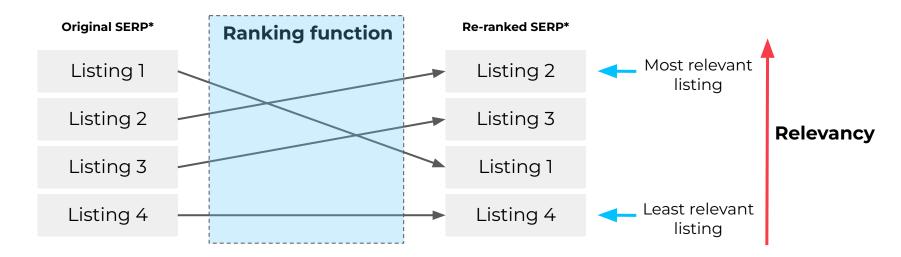
- Classic search infrastructure and workflow; no "easy hooks" for AI
- Latency budget: 10's of ms
- User search experience at all costs







Phase 1: use ML to re-rank search results



We want to re-rank the search results so that **more relevant** listings are placed **higher**.



MLOps

What is it and why do we even care?



What

- ML(Dev)Ops
- "Set of practices that aim to deploy and maintain ML models in production reliably and efficiently^[1]"

[1] S. Shankar, R. Garcia, J. M. Hellerstein, and A. G. Parameswaran, "Operationalizing machine learning: An interview study," arXiv preprint arXiv:2209.09125, 2022.





• ML application development, deployment, and maintenance challenging





- ML application development, deployment, and maintenance challenging
- Variance between data, use-cases, constraints, ...





- ML application development, deployment, and maintenance challenging
- Variance between data, use-cases, constraints, ...
- No universal solutions





- ML application development, deployment, and maintenance challenging
- Variance between data, use-cases, constraints, ...
- No universal solutions
- MLOps still nascent







• Al as an implementation detail





- Al as an implementation detail
- Use-case-driven "MLOps"





- Al as an implementation detail
- Use-case-driven "MLOps"
- Iterate on bottlenecks and requirements





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- Judicious resource allocation





- Al as an implementation detail
- Use-case-driven "MLOps"
- Iterate on bottlenecks and requirements
- Good feedback signals
- Judicious resource allocation
- Starting small but soon





Data Pipelines

In 5 minutes or less



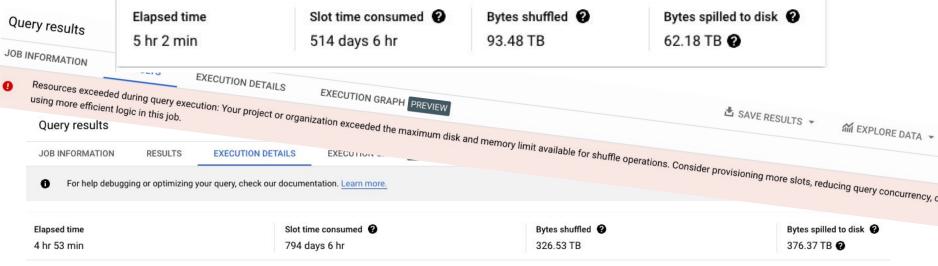
Data Pipelines...?

- A dozen 500+ line SQL files
- Manually executed



Data Pipelines...?

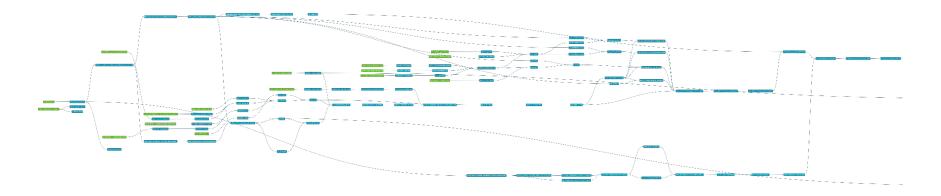
- A dozen 500+ line SQL files
- Manually executed
- Painful





Data Pipelines...

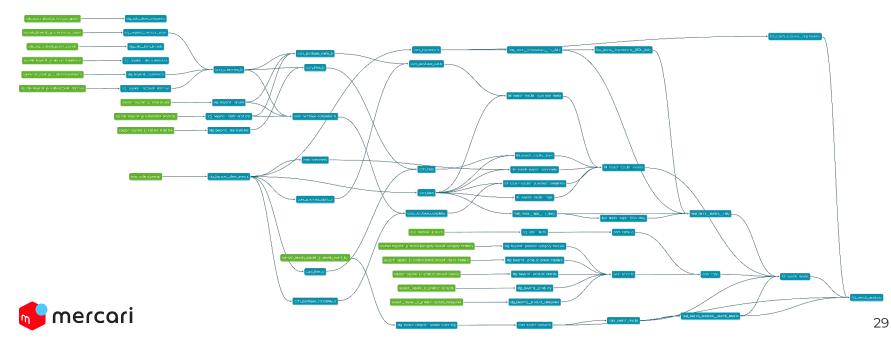
- A hundred 10-100+ line SQL files
- Inefficient & inaccessible
- Painful





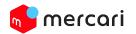
Data Pipelines!

- A few dozen SQL files
- Structured
- Painful?



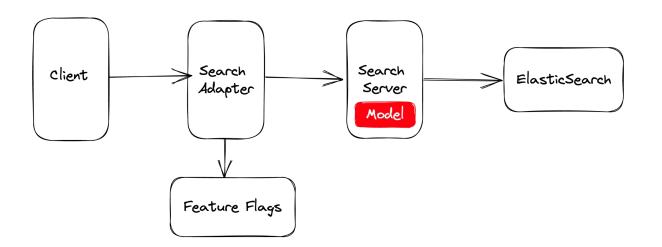
System Evolution

Growing an ML system while running the business



v0: In-Situ Model Serving

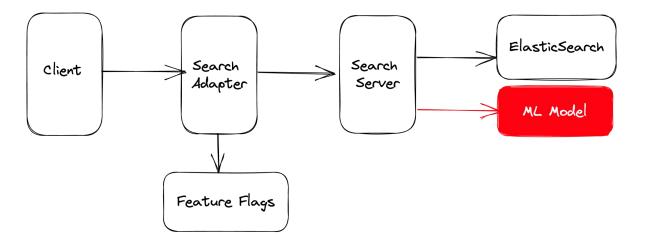
- Model serving within search server
- Features computed in search workflow





v1: Decoupled Model Serving

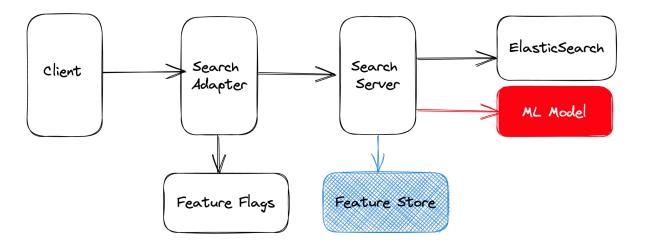
- Custom-made python microservice for model serving
- RPC with timeout and "baseline" response
- Basic monitoring of production metrics





v2: Simple Feature Store

- Offline feature store from data pipelines
- Online feature store with direct ETLs
- Timeouts & failsafes redux





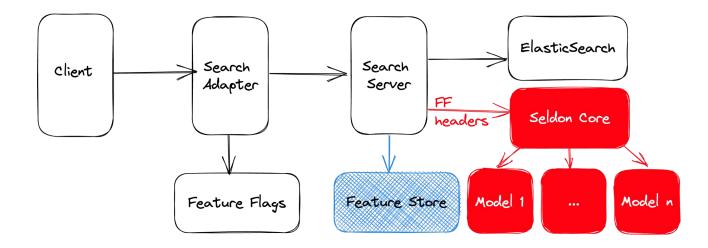
A/B Test Setup: Before

Iprod] Remove frequent update auth Secret in mercari-searchx-jp size/XS #60568 by TeoZosa was merged on Jan 23 · Approved	٥	5
Iprod] Remove frequent update Secret reference in x-server env size/M #60567 by TeoZosa was merged on Jan 23 · Approved	٥	() 11
Revert "[prod] Remove frequent update ETL Memorystore Secret from x-server env" size/M #60434 by TeoZosa was merged on Jan 18 · Approved	0	1 4
[dev] Remove tokenization LtR server resources size/L #60405 by TeoZosa was merged on Jan 19 - Approved	6	₽ 2
[dev] Remove IU frequent update LtR service resources size/L #60310 by TeoZosa was merged on Jan 17 • Approved	٢	Ç 4
[prod] Remove IU frequent update LtR service resources size/L #60307 by TeoZosa was merged on Jan 17 · Approved	٢	Γ 6
[dev] Remove frequent update ETL Memorystore Secret from x-server env < size/XS #60294 by TeoZosa was merged on Jan 16 · Approved	٢	7
[prod] Remove frequent update ETL Memorystore Secret from x-server env size/M #60291 by TeoZosa was merged on Jan 17 · Approved	٢	₽ 24
\$~ [dev] Remove deprecated A/B test feature flag: TW0-12251_more_frequent_updates size/XS #60289 by TeoZosa was merged on Jan 16 • Review required	٢	Д 3
[prod] Remove deprecated A/B test feature flag: TW0-12251_more_frequent_updates size/M #60288 by TeoZosa was merged on Jan 17 • Review required	٢	₽ 26
[dev] Remove deprecated A/B test feature flag: TW0-12250-per_item_stats_tokenization size/XS #60287 by TeoZosa was merged on Jan 17 • Review required	٢	Ç 17



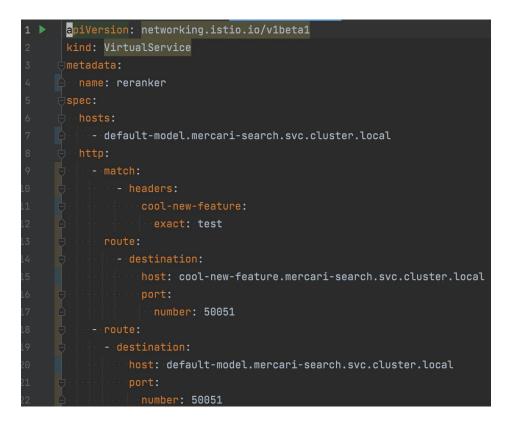
v3: Batteries-Included Model Serving Framework

• Seldon & Istio





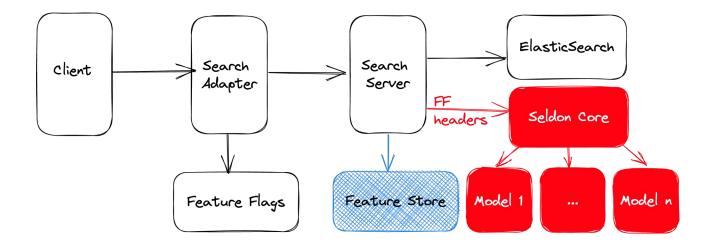
A/B Test Setup: After





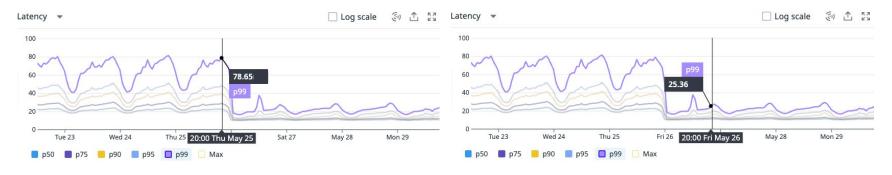
v3: Batteries-Included Model Serving Framework

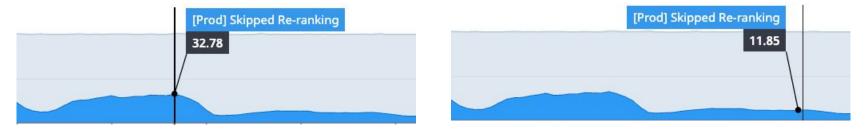
- Seldon + Istio
- Shadow Traffic





Shadow Traffic: Test in Prod

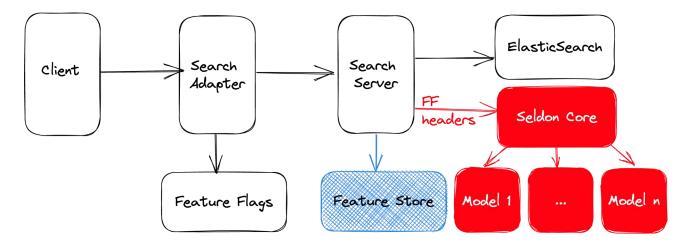






v3: Batteries-Included Model Serving Framework

- Seldon + Istio
- Shadow Traffic
- Fine-grained model serving





Future Directions

Reliability & Effectiveness



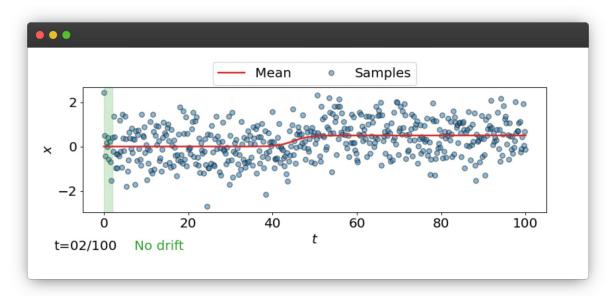
Monitoring



Outlier, adversarial, and drift detection



Drift Detection



Detect **drift** to preempt downstream performance degradations



Conclusion

- ML-enhanced search possible with incremental investments
- Resilience of use-case-driven platforms and systems
- Engineering/business trade-offs
- "One is too small a number to achieve greatness"
- Build something meaningful by building together



