Streaming datasets for Personalization

Shriya Arora

Senior Data Engineer Personalization Analytics



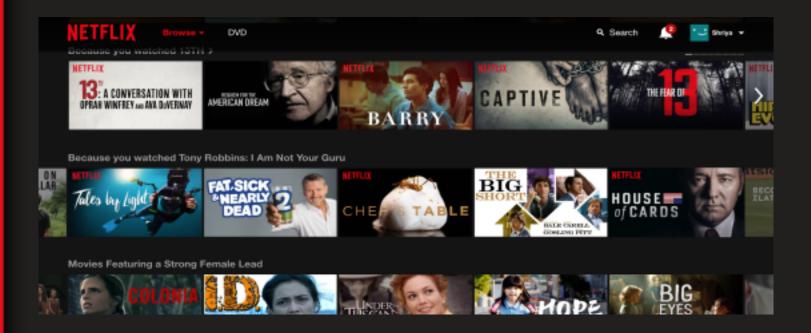
What is Netflix's Mission?

Entertaining you by allowing you to stream content anywhere, anytime



What is Netflix's Mission?

Entertaining you by allowing you to stream personalized content anywhere, anytime



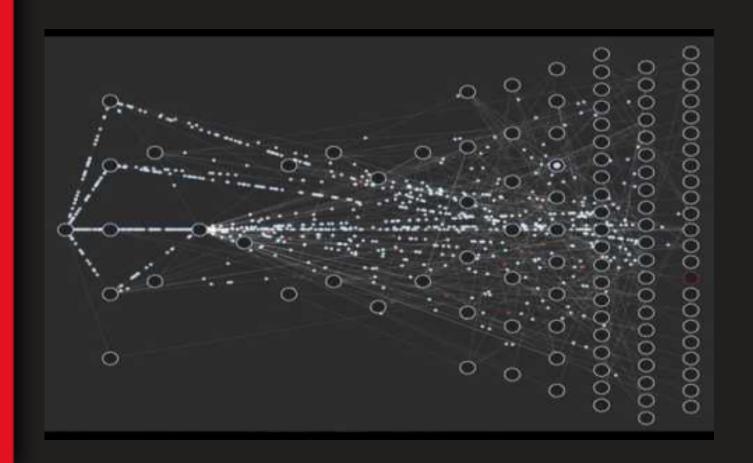


How much data do we process to have a personalized Netflix for everyone?

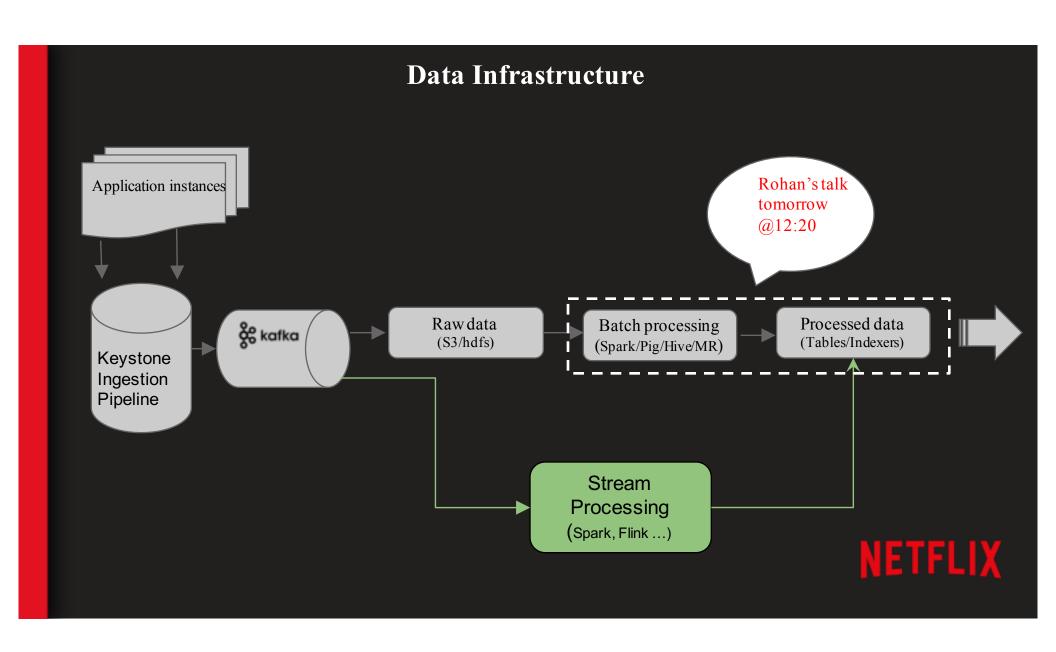
- 93M+ active members
- 125M hours/day
- 190 countries with unique catalogs
- 450B unique events/day
- 600+ Kafka topics



A SERIES OF PLAYBACK EVENTS



NETFLIX



Our problem: Using user plays for feature generation, discovery, clustering ..



Why have data later when you can have it now?

Business Wins

- Algorithms can be trained with the latest + greatest data
- Enhances research
- Creates opportunity for new types of algorithms

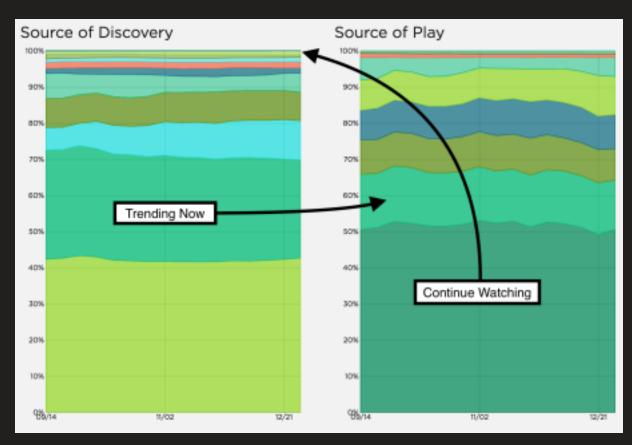
• Technical Wins

- Save on storage costs
- Avoid long running jobs



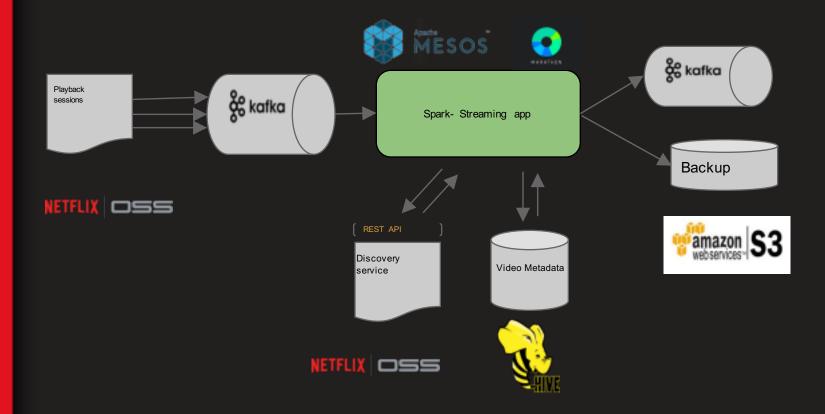
Source of Discovery / Source of Play





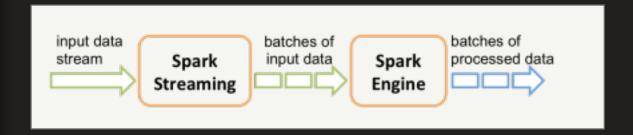


Source-of-Discovery pipeline





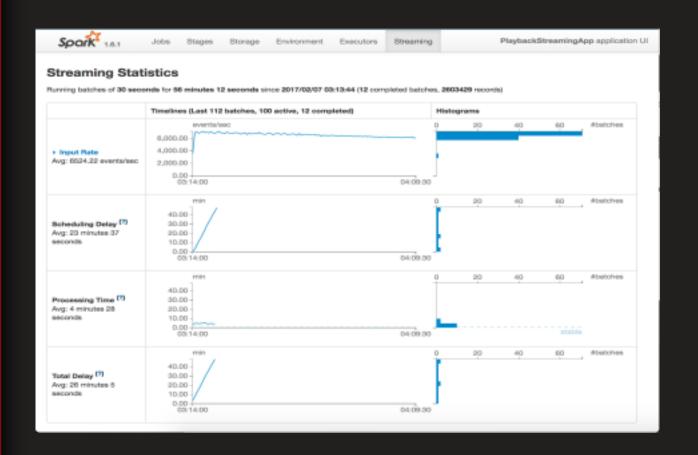
Spark Streaming



- Needs a StreamingContext and a batch duration
- Data received in DStreams, which are easily converted to RDDs
- Support all fundamental RDD transformations and operations
- Time-based windowing
- Checkpointing support for resilience to failures
- Deployment



Performance tuning your Spark streaming application



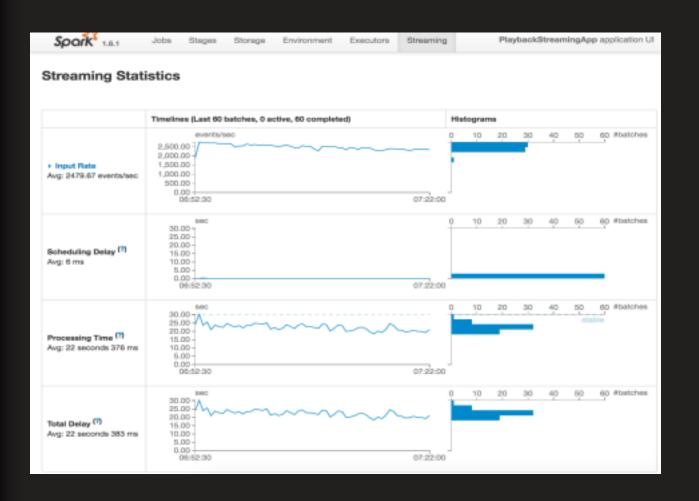


Performance tuning your Spark streaming application

- Choice of micro-batch interval
 - The most important parameter
- Cluster memory
 - Large batch intervals need more memory
- Parallelism
 - o DStreams naturally partitioned to Kafka partitions
 - Repartition can help with increased parallelism at the cost of shuffle
- # of CPUs
 - <= number of tasks
 - o Depends on how computationally intensive your processing is



Performance tuning your Spark streaming application





Challenges with Spark

- Not a 'pure' event streaming system
 - Minimum latency of batch interval
 - o Un-intuitive to design for a stream-only world
- Choice of batch interval is a little too critical
 - o Everything can go wrong, if you choose this wrong
 - o Build-up of scheduling delay can lead to data loss
- Only time-based windowing*
 - Cannot be used to solve session-stitching use cases, or trigger based event aggregations

* I used 1.6.1





Questions?

Stay in touch!



@NetflixData

