Knowledge from Nothing: Creating Data Pipelines for User Insights from Zero

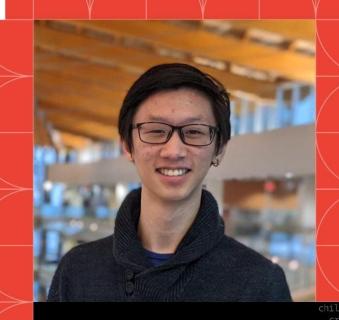
Jeffrey Leung

Software Developer









Soogle Developer Groups

devfest

Vancouver



Knowledge from Nothing Creating Data Pipelines for User Insights from Zero

JEFFREY LEUNG





The exponential growth of data is undisputed, but the numbers behind this explosion - fuelled by internet of things and the use of connected devices - are hard to comprehend, particularly when looked at in the context of one day



every day

billion emails are sent



of data created by Facebook, including

350m photos

100m hours of video watch time

DEMYSTIFIYING DATA UNITS

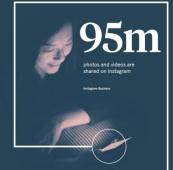
From the more familiar 'bit' or 'megabyte', larger units of measurement are more frequently

Unit	Value	Size
	0 or 1	1/8 of a byte
B byte	8 bits	1 byte
KB kilobyte	1,000 bytes	1,000 bytes
MB megabyte	1,000² bytes	1,000,000 bytes
GB gigabyte	1,000 ⁵ bytes	1,000,000,000 bytes
TB terabyte	1,000° bytes	1,000,000,000,000 bytes
PB petabyte	1,000 ⁵ bytes	1,000,000,000,000,000 bytes
EB exabyte	1,000° bytes	1,000,000,000,000,000,000 bytes
ZB zettabyte	1,000 ⁷ bytes	1,000,000,000,000,000,000,000 bytes
VB vottabyte	1.000* bytes	1,000,000,000,000,000,000,000,000 bytes

"A lowercase "b" is used as an abbreviation for bits, while an uppercase "B" regresents bytes



of data will be created every day by 2025





3.5bn



3.9bn

320bn

306bn emails to be sent each day by 2020

emails to be sent

each day by 2021

ACCUMULATED DIGITAL UNIVERSE OF DATA 4.4ZB **44ZB**

Searches made a day

Searches made

a day from Google

Why do you(r leaders) want data?

Analyze past performance.

Understand current performance.

Experiment on new features.

Make informed business decisions.

Machine Learning for new insights.



Outline

- 0. data engineering fundamentals
- 1. starter tools
- 2. meaning from a mess
- 3. organizational maturity



0. the fundamentals



Where does it come from, where does it go?



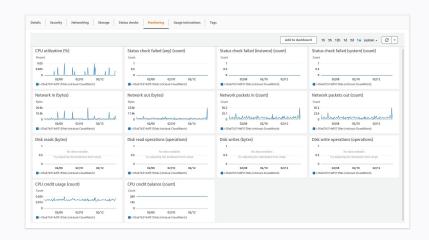


What kind of data?

User behaviour (e.g. signups, interactions)



Operational metrics (e.g. cpu, memory, latency, network)







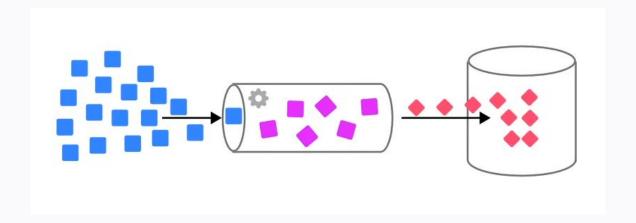


Streaming Data Pipeline

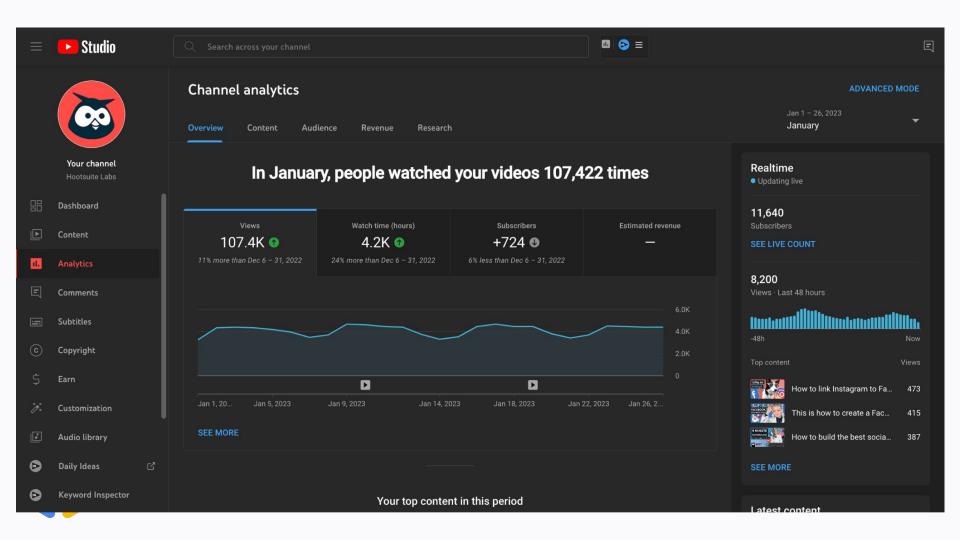




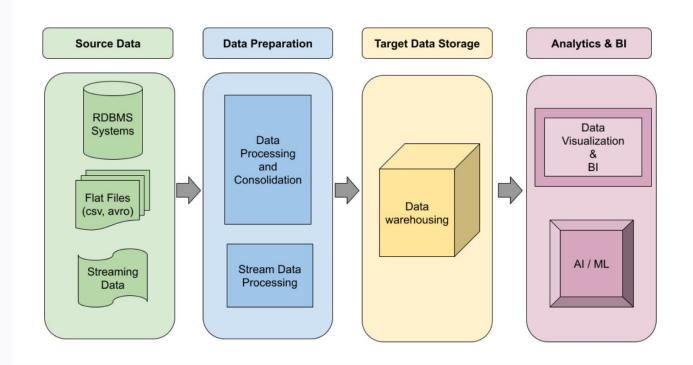
Streaming Data Pipeline





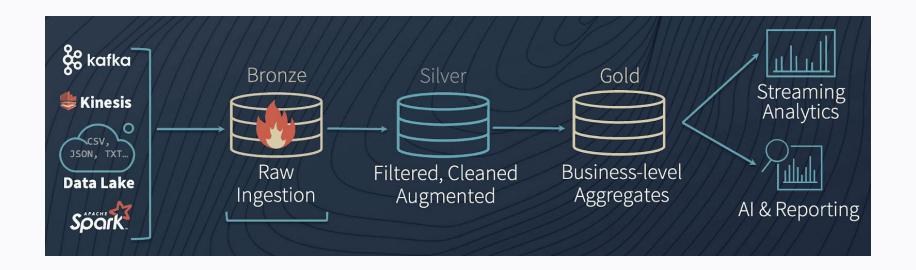


Batch Data Pipeline



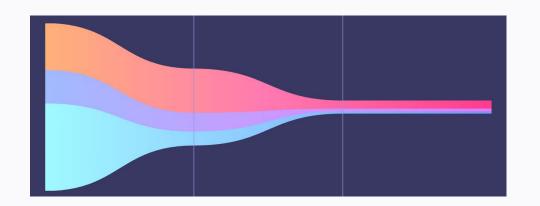


Data Pipeline Architecture





The Dataflow



01. ingestion and storage

pseudonymization anonymization replayability verification from source

02. cleaning and filtering

sanity checks standardization elimination of invalid data

03. aggregating and analyzing

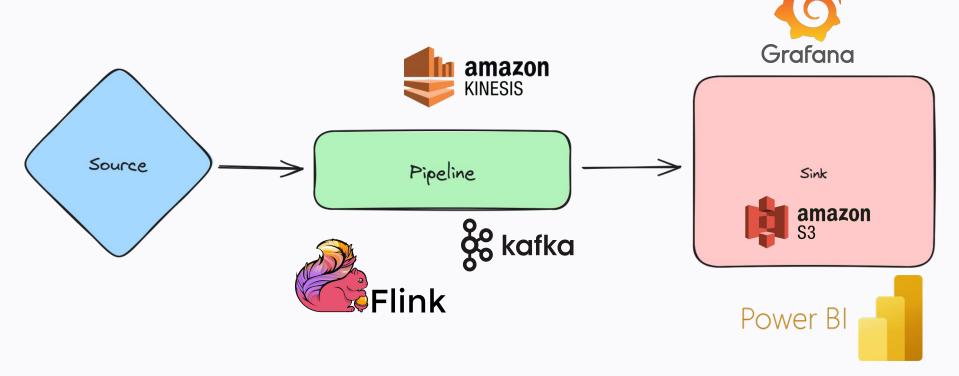
averaging
determining patterns
alerting
p50/p99
diffs across days, weeks,
months, or years
graphing time series data
predictive analysis



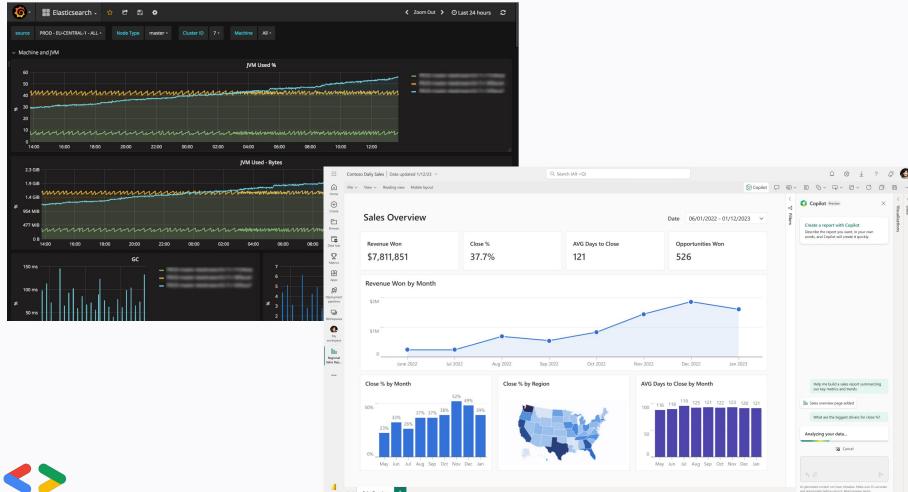
1. the tools



Tools for Streaming

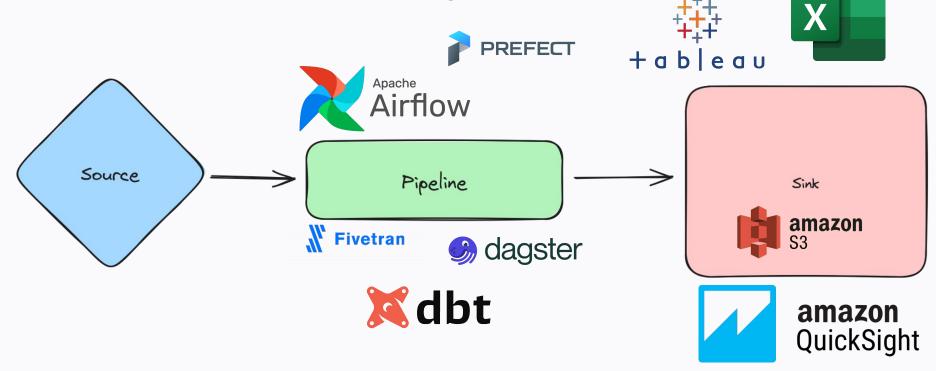








Tools for Batch Processing





Data Warehouses











Something Else: Mixpanel







2. the why



Who wants this data?

How do they want to consume it?

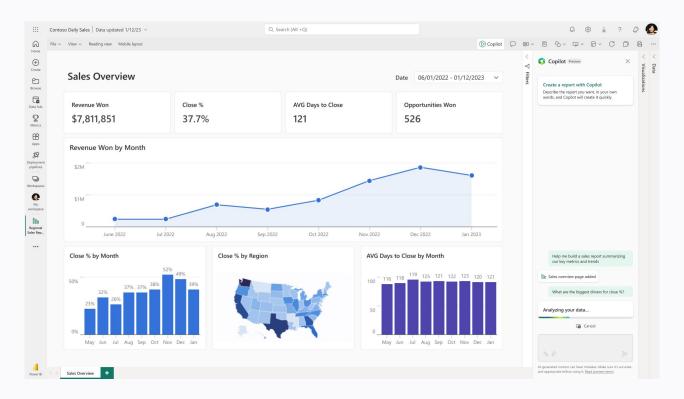
What decisions do they want to make with it?

What is this data intended to measure?

How do we measure what we actually want data on?



Align on expected output.





What **business value** does this data provide?

Is this the best way to provide it?



What I does t

Is this the best v





Data is your **Product**.

Understand the requirements.

Validate the user need.

Design an MVP.



Present clearly.

Use graphs and charts.

Compare for context.

p50, p90, p99.



3. data org maturity



How to draw an Owl.

"A fun and creative guide for beginners"

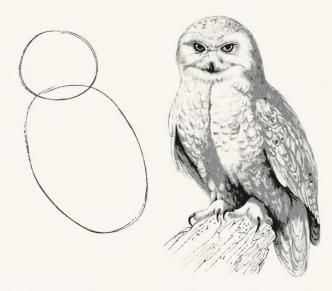


Fig 1. Draw two circles

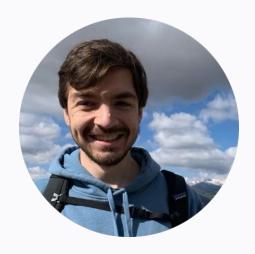
Fig 2. Draw the rest of the damn Owl



Follow these people



Zach Wilson
Staff Data Engineer
Prev. Facebook, Netflix, Airbnb



Benjamin Rogojan

Seattle Data Guy

Prev. Facebook



Get that data integrity

Compliance (GDPR, DSAR, COPPA, DMA) - anonymization, retention, deletion

Quality verification (e.g. Great Expectations, deequ)

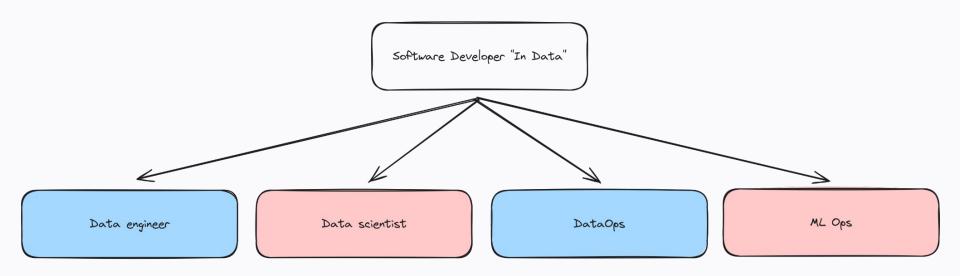
Documentation (e.g. Amundsen)

Monitoring and alerting





Evolution of a data-driven organization





Evolution of a data-driven organization

Experimentation.

Long-range performance analytics.

Deep learning for understanding/prediction.

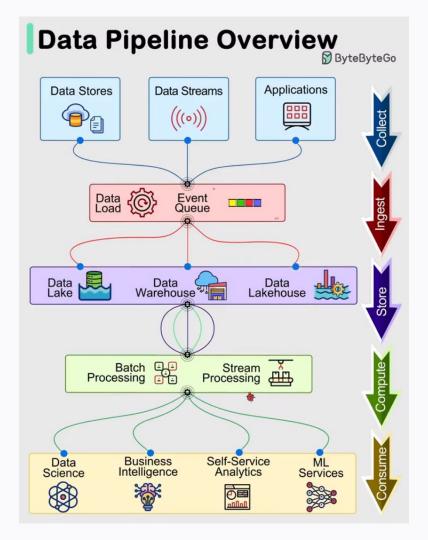


Questions?



GitHub: jleung51







The Dataflow

01. ingestion and storage

retention: short

02. cleaning and filtering

retention: medium

03. aggregating and analyzing

retention: long

