

Data Council 2024 From Twilio to Propel: Building Real-Time Customer-Facing Analytics at Scale

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# l'm Nico Acosta

- Computer Scientist.
- Co-founder and CEO at Propel.
- I was employee 65 at Twilio. Spent a decade building API, data, and AI products.
- AWS Global expansion.
- Co-founder of Masivian.







#### Part 1

Building customer-facing analytics at Twilio



- Comms Platform as a Service
- SMS, Voice, and Email APIs
- ✓ Public company. IPO'd in 2016
- > 3 TB of data per day
- > 4 billion raw events per day

#### Part 2

Lessons learned and building Propel

Propel

- Analytics Platform as a Service
- Customer-facing analytics
- APIs, SQL, Embedded & Portals
- Built to handle insane scale
- Completely Serverless

#### Part 3

Live coding: Real-time customer-facing analytics in 5 mins



Unify data from multiple sources

- Transform in real-time
- Data-Serving APIs and SQL
- Build UI with React Components
- Multi-tenant access controls.

# Customer-Facing Analytics



#### **Customer-Facing Analytics**

Customer-facing analytics are insights provided directly to you, <u>the customer</u>, about the service or product you are using.

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2 Elliott Smith	2 Lovesong - Re
3 Radiohead	3 Between The B
4 Jeff Buckley	4 Lover, You Sho
5 Fiona Apple	5 Grace
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104,974	Rock

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Spotify

# What do the <u>best</u> Customer-Facing Analytics experiences have in common?



# Customers are willing to pay for them

# What does it take to build an analytics product that customers <u>are willing to pay for</u>?

Sub-second query latencies Sub-minute data freshness Consume data multiple ways

#### **Customer-Facing Analytics at Twilio**

- Processing over **3 TB** per day
- Over **5 billion** raw events per day
- Every call emitted an event every 10 seconds
- All processed in real time
- Powering mission critical applications

#### **Mission Critical Use Cases @ Twilio**

- Voice Insights: Real time insights for debugging.
- Fraud detection: Direct and indirect fraud.
- **Carrier routing:** Quality and cost optimization.

# A lot of things can (and do) go wrong When operating communications at a global scale

# **Voice Insights at Twilio**

#### High-value customers had unique reporting requirements.

Simple, in-product dashboards didn't cut it

#### Ad-hoc and recurring data requests.

They were a poor customer experience and labor intensive to fulfil

High-value customers needed to consume data their way.

Via embedded dashboards in the Console, SQL, BI Tools, or APIs.



#### Gave high-value customers the confidence to scale on the platform

Scaled the voice product from \$10M ARR to \$100ARR in the next 3 years.

#### Freed-up teams to focus on the core job.

Eliminated all ad-hoc data requests.

#### Became one of the fastest revenue-generating products

60% net-dollar expansion on adopters

5 billion events per day



Sub-second query latencies

Lots of dimensions and time zones



Sub-minute data freshness

Really expensive to run





#### **Productizing Analytics at scale:** The Fundamental Trade-off



# **Productizing Analytics at scale:** The Fundamental Trade-off



Decisions we made (not all of them ideal):

- Aggregate in UTC
- Key Performance Metrics:
  - Duration, pdd, and jitter.
- Aggregations:
  - Min, Max, Avg, 95th Percentile, 99th Percentile
- Dimensions:
  - Predefined set of dimensions

# Twilio Voice Insights Architecture



#### Twilio Voice Insights Data Contracts

Data quality is the responsibility of the producer

#### • Time:

- Source timestamp can't be always trusted.
- ALWAYS log the time you receive the event
- ID is idempotent , i.e. the dedupe key
- Extensible to new metrics
- Extensible to new dimensions

#### Twilio Voice Insights Data Contracts

Data quality is the responsibility of the producer

message CallEvent {
 // when the event occured utc
 optional uint64 event\_time
 // when we received said event utc \*\*\*\*
 optional uint64 logged\_event\_time
 // unique identifier for said event - =
 optional string event\_id
 // encapsulates a logical end to end re
 optional string route\_id
 // used to store multiple types in this
 optional CallEventType event\_type
 // when event type is signaling\_event
 // use with the event\_type and logical
 optional Dimensions event\_dimensions
}

message SignalingEven	t {		
optional string nam	e	=	1;
optional SignalingE	ventType event_type	=	2;
optional PddEvent p	dd	=	3;
optional CallStateE	vent call_state	=	4;
}			
enum SignalingEventTy	pe {		
unknown_event_type	= 0;		
pdd	= 1;		
call_state	= 2;		
}			
<pre>message PddEvent {</pre>			
optional float pdd	= 1;		
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m	essage Met	tricAggregation {		
	optional	string metric	=	1;
	optional	<pre>uint64 window_start</pre>	=	2;
	optional	uint64 window_end	=	3;
	optional	<pre>string window_interval</pre>	=	4;
	optional	uint32 samples	=	5;
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	optional	Dimensions dimensions	=	8;
	optional	string dimension_hash	=	10;
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#### From a Product to a Platform

Lessons learn at Twilio and how we applied them to Propel

# Most SaaS Companies Spend a Lot of Effort & Get Little Value



Propel

# **Customer-Facing Analytics Roadmap**

The highest-value, low effort products are not always obvious.

	Who It Serves	Engineering Effort	Revenue Opportunity
1. SQL Connector	High-value customers	Low	High
2. White Label Reporting Portal	High-value customers	Low	High
3. Analytics APIs	High-value customers	Medium	High
4. Embedded Dashboards	Long-tail customers	Medium	Low
5. Custom-built In-product Dashboards	Long-tail customers	High	Low

### From an Analytics Product at Twilio To a Platform at Propel



#### Propel

# **Propel** Architecture



## **Propel** Architecture





# See it live

TacoSoft, our imaginary SaaS company, ships its first real-time analytics product

- → Ingest data from our data warehouse
- → Ingest streaming data from Postgres-Kafka
- → Transform and enrich it in real time
- → Set multi-tenant access controls
- → Expose it via the API
- → Build it into a customer-facing dashboard

#### **Demo recap**

#### • Ingest

- Ingested data from Snowflake
- Ingested data from Kafka

#### • Transform

- Flattened JSON from Kafka Messages
- Enriched data joining datasets

#### • Serve

- Served it via Analytics APIs
- Types of APIs: Time Series

#### • Visualize

- Custom dashboard in React
- Using the Propel React components



# If you remember 3 things from this talk:

- 1. Best analytics products are those that customers are **willing to pay for**.
- 2. To successfully productize analytics, iterate quickly the **Cost-Effectiveness**, **Performance**, and **Flexibility** trade-off.
- 3. The highest-value, low effort products are not always obvious.

# **Propel** is now generally available!



# www.propeldata.com





Turn customer data requests into revenue-generating analytics products

Propel

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