Real-Time Deep-Link Analytics for Big Graphs

Challenges and Solutions

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BigGraph Meetup October 11, 2017

Welcome to



We meet to network, share, discuss, and invent together graph technologies that empower the analysis capabilities needed for today's most critical enterprise applications.

Goal: real-time and mutable Big Graph data management platform supporting deep link analytics

• Big Graph? Billion-scale Graphs, with 10B to 100B+ vertices

and/or edges.

Real-time milli-second query response time on Big Graphs.

Mutable data mgmt supporting real-life scenarios, > 100K+

updates per second.

Deep link analytics queries which traverse 3 to 10+ hops deep

into the graph.

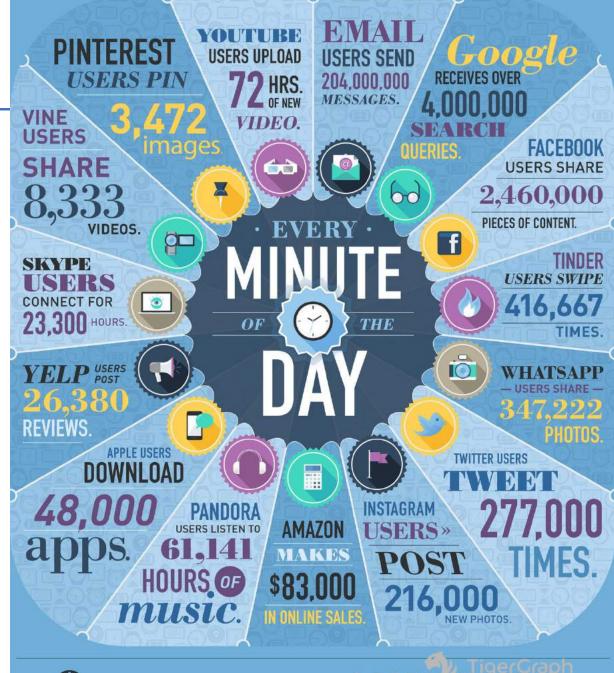


The Data Deluge

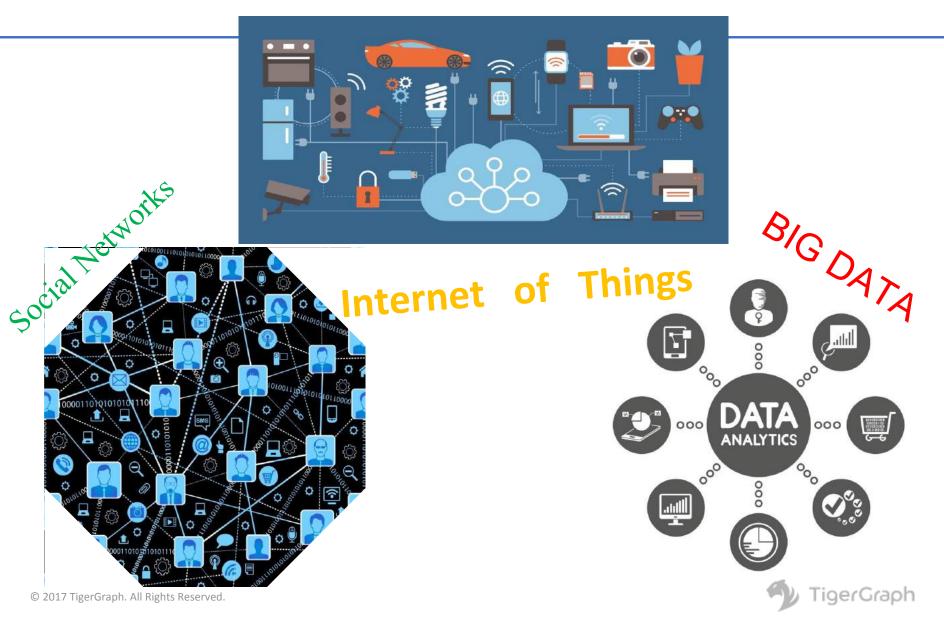
Data Never Sleeps 2.0

from Domo.com

https://www.domo.com/learn/data-never-s leeps-2

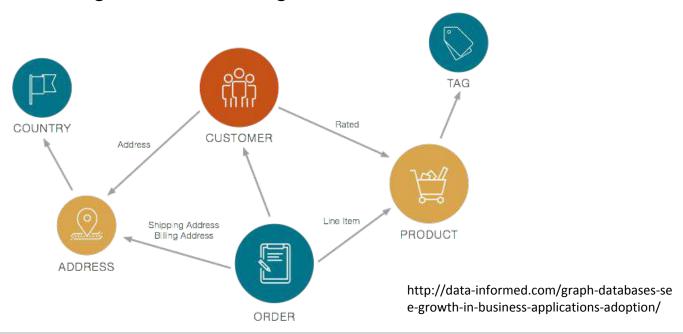


The Networked Age



Graph Data Analytics to the Rescue

- Natural storage model for connectional data
- Natural model for many types of transactions
- Natural computational model for knowledge/inference/learning chaining and combining observations





Challenge: Scalability

- Datasets are getting bigger
- Outgrowing a single server in many cases.
- More data → better models/predictions/results
- Problem: Some graph DBs cannot distribute across multiple nodes.

Requirement:

Data Platform must scale-up and scale-out well

X axis: Year Y axis: Global data Unit of measurement: Zettabytes 1ZB = 1,000° bytes 25 20 15

DATA GROWTH

Note: Post-2013 figures are predicted. Source: UNECE

https://www.theneweconomy.com/strategy/big-data-is-not-without-its-problems

2013

2014



2018

Challenge: Real-Time Processing

Many applications require real-time processing. Gets harder as data sizes get bigger.



- Processing transactions
- Engaging customers
- Catching fraud in progress
- Managing live systems
- Ingesting streaming data

Requirement:

Speed needed both for querying and data updates/loading



Challenge: Deep Link Analytics

- Each additional hop in a graph reveals more information / knowledge / evidence
- Deeper traversal → better models/predictions/results
- Problem: Most graph DBs slow down or timeout after 2 hops.

Requirement:

Data Platform must traverse multiple hops efficiently





Need: Scalable Real-Time Deep Link Analytics

Deep Link Analytics: Queries With 3 Or More Hops



Deep Link Analytics are essential for enterprise applications including Al and machine learning apps

fraud and risk management, anti-money laundering personalized recommendation, knowledge graph, customer identity graph, supply-chain logistics, ...



Requires a real-time graph platform that shatters current graph database speed and performance limitations



Solution: Native Parallel Graph

TigerGraph: The First Native Parallel Graph

Designed From the Ground Up For Computational Parallelism







Native Parallel Graph

Automatic Computational Parallelism



Each vertex/edge is not only a unit of storage, but is also a computational unit



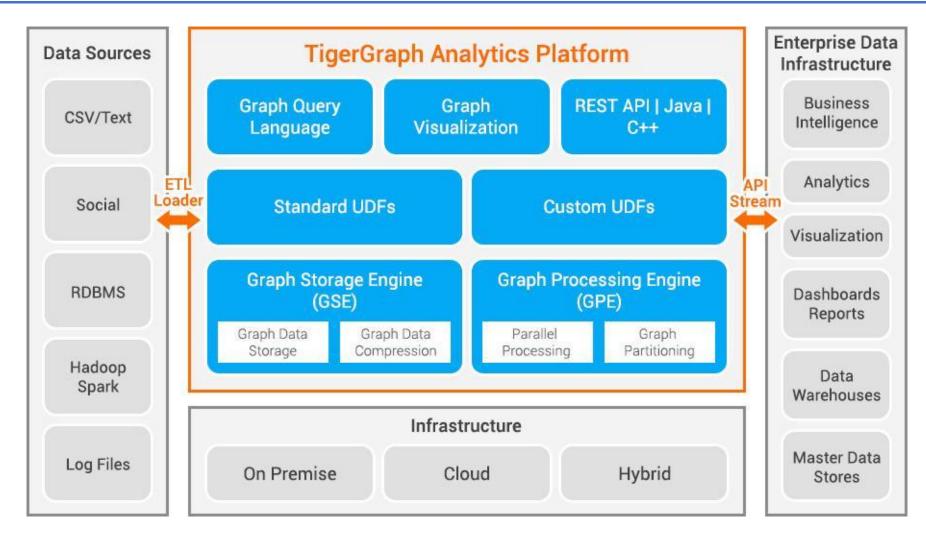
Each vertex/edge is processed in parallel



TigerGraph engine will automatically scale the computation across all threads and CPU cores available



Real-Time Graph Analytics Platform



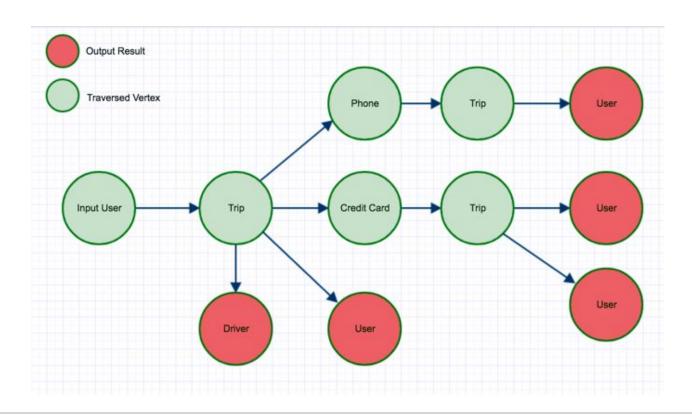
Demo: Queries Per Second on a Big Graph

- Graph Schema and Size: Delivery service
 - 7 vertex types: Order, User, Location, CreditCard, etc.
 - 9 edge types: Order_User, Order_Location, Order_Card, etc.
 - 3.4 Billion vertices
 - 22.5 Billion edges
- Test Server
 - Amazon EC2 i3.16xlarge
 - 64 vCPU, 488 GiB Memory, 15 TB SSD Storage



Demo: Queries Per Second on a Big Graph

Starting with a Driver_id X, find all other Driver_ids/User_ids connected to the Driver_id X through phone_number, credit_card connected via trips in 4 steps

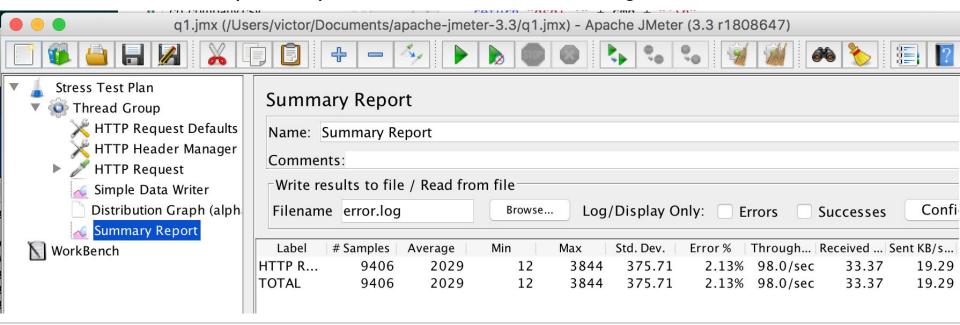




Demo: Queries Per Second on a Big Graph

Based on the graph density, each query can traverse about 100K edges. The live test shows that we can process 98.0 queries per second per machine \rightarrow 10M edges per second, for a 4-hop query on a graph with 22B edges.

Even better speed is possible for a less demanding test.



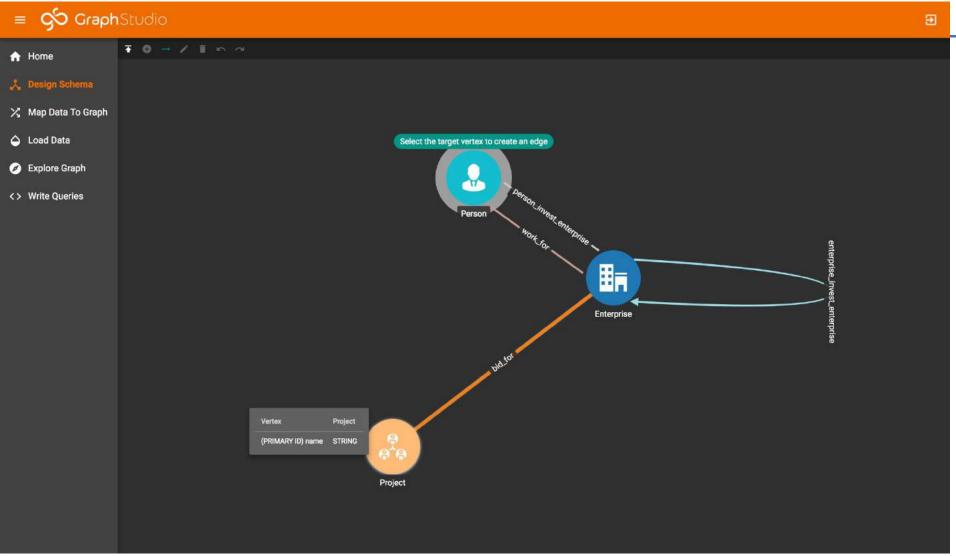
A Closer Look At the TigerGraph System



web browser based tool to visually and interactively define, create, explore and query big graphs

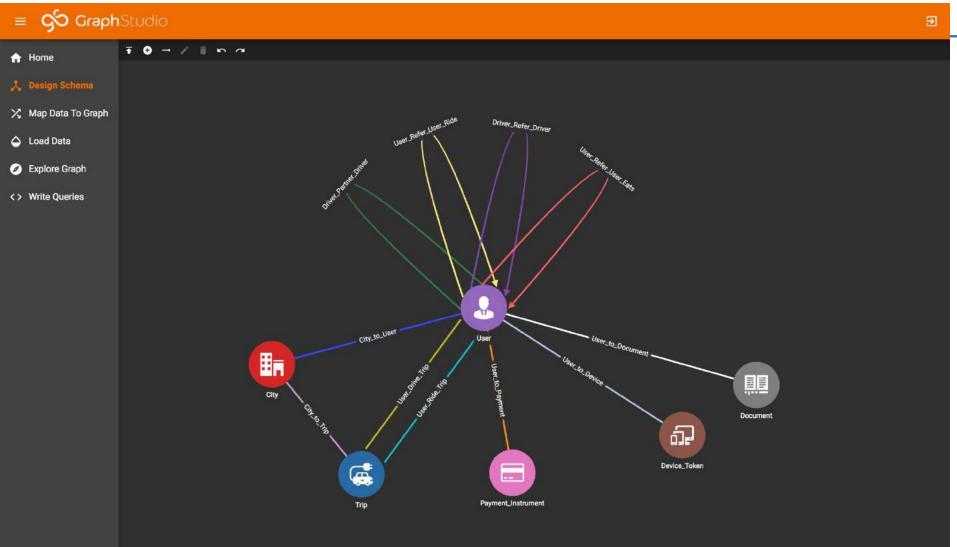


GraphStudio: Visual Graph Schema Designer





GraphStudio: Visual Graph Schema Designer



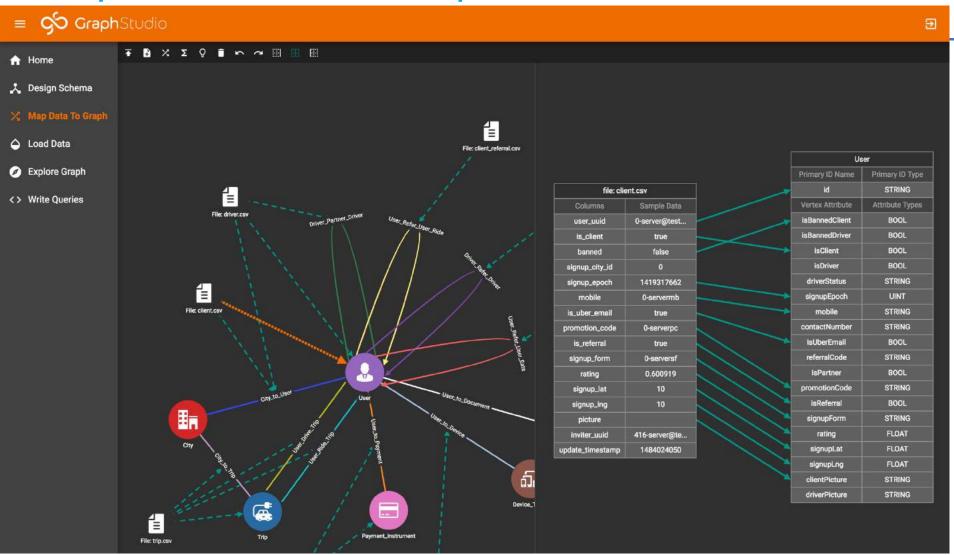


GraphStudio: Visual Graph Data Loader

≡ 🕉 GraphStudio FBX Z Q I n a H H H ↑ Home Design Schema △ Load Data Explore Graph Write Queries a Œ File: trip.csv



GraphStudio: Visual Graph Data Loader





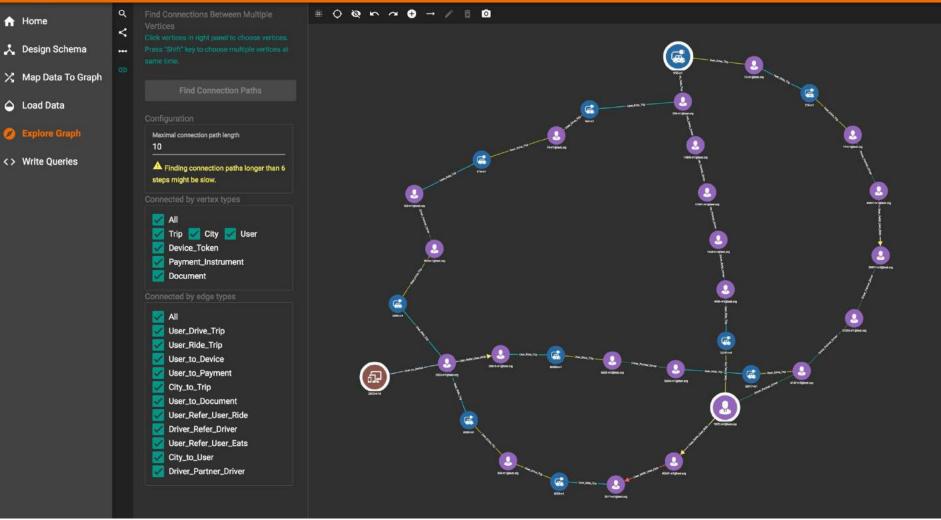
GraphStudio: Graph Explorer

≡ S GraphStudio ♠ Home Enter vertex id 🏃 Design Schema Enter a number Pick Vertices 🔀 Map Data To Graph △ Load Data Enterprise Person <> Write Queries



GraphStudio: Graph Explorer

= \$ GraphStudio





GraphStudio: Graph Query Editor

≡ 🕉 GraphStudio **GSQL Queries** 日日子 > 日 ♠ Home T = CREATE QUERY MultiTrip (VERTEx<Trip> trip, int step, int startDate=0, int endDate=1999999999) FOR GRAPH **Enter Query Parameters** poc_graph { Design Schema SetAccum<VERTEX> @@tripSet; X Map Data To Graph OrAccum<bool> @fromDriver, @fromRider; 2000000000 SetAccum<VERTEX> @@userSet; SetAccum<EDGE> @@edgeSet; △ Load Data Start (ANY) = {trip}; Explore Graph 10 Start = SELECT t FROM Start:s-((User_Drive_Trip|User_Ride_Trip):e)-:t Write Queries case when e.type == "User_Drive_Trip" then t.@fromDriver += true else t.@fromRider += true ,@@userSet += t ,@@edgeSet += e **Run Query** Ê



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- Senior Technical Product Manager

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See all our open positions at www.tigergraph.com/join-us/



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More questions?

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