

Introduction to Graph Analytics



Advanced Analytics & ML for Connected Data



Graph Is How WE THINK



Modern businesses need intelligent answers to complex questions...



Graph Stores Data Differently



Relational databases don't tell the whole story

Relational databases store facts in tables

Name	Location	Product	Bank	•
John	Palm Springs	iPhone	JPM Chase	٠
David	Los Angeles	MacBook	Capital One	•
Lisa	San Francisco	Watch	JPM Chase	٠
Jim	Palm Springs	iPad	Capital One	•
Sally	San Francisco	Watch	HSBC	٠
Steve	Los Angeles	iPhone	JPM Chase	•

- Cannot easily model indirect relationships
- Cannot run queries across data sets without slow joins
- Cannot add new relationships without schema changes

Surface level sampling



- Flips the perspective from facts to relationships
- Scales massively without sampling
- Faster queries at greater depth

Rich, intuitive analysis

Graph databases make AI/ML models faster and more accurate



Richer, smarter data

Relationships-as-data

• Connect different datasets, break down silos

Deeper, complex questions

Accelerated performance

Explainable results

- Look for semantic patterns (implicit relationships)
- Easily and quickly search far and wide
- High-speed queries
- Relationship powered algorithms and machine learning
- Intuitive models, queries and answers
- Visual exploration and results

% of data and analytics innovations using graph technologies



Machine Learning and Algorithm Support

(70+ Github)



Algorithm Types

Centrality

Assign numbers or rankings to each vertex corresponding to their network position

Classification

Classify the vertices into sets according to some external rule

Community

Group the vertices so that each group is densely connected

GraphML/Embeddings

Convert the neighborhood topology of each vertex into a fixed size vector of decimal values

📮 🛛 Path

Find the best paths from one vertex to another (shortest, lowest weight, or other criteria)

Similarity

Compute similarity between pairs of items

Topological Link Prediction

Predict the existence of a link between two entities in a network

Frequent Pattern Mining

Find subgraph patterns that occur the most frequently

Common Use Cases

Uses cases to apply graph include:

- Entity Resolution
- Customer Journey / C360
- Recommendation Engines
- Supply Chain
- Fraud/AML
- Data Fabric
- Cybersecurity
- AI & Machine Learning





Q&A Break

Focus on Fraud



Tree of Pain - Use Cases



Visualization - Functionality by User

TG STUDIO







Back-End

≡ 🕉 GraphStudio (i) e Admin Q 合 Home 🙏 Design Schema 🔀 Map Data To Graph Vertex 1 Person Paul Load Data Vertex 2 Person Robert Explore Graph Vertex 3 Company Vibber <> Write Queries Vertex 4 Company VibberTorch Vertex 5 Project Umbrella Telecom Fraud Detection . Find Connection Paths WePipe Movie Recomm . VibberTorc BidFor Query endpoint Maximal connection path length 6 Umbrella Telecom Fraud Detec NOTE: Please make sure your firewall settings allow access to port 9000. http://localhost:9000/query/demo_fc_graph/generic_one_hop Person 🔽 Company Project CREATE QUERY generic_one_hop() FOR GRAPH VorkFor PRINT "generic_one_hop works!"; PersonInvestCompany 4 } CompanyInvestCompany J BidFor

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Front-End

Financial Crimes Workbench Demo

- Rules / Alerts
- Dashboards
- Analytics
- Case & Alert Mgt





Q&A Break

Get in touch with us!



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Resources

- <u>TigerGraph Fact Sheet</u>
- Start for Free with TGCloud
- <u>TigerGraph Blog and Resource Hub</u>
- <u>TigerGraph Demo Library</u>
- <u>Analyst Reports</u>