



Driving Ecommerce Profits with Personalized Recommendations using Machine Learning and Graph Analytics

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About Us



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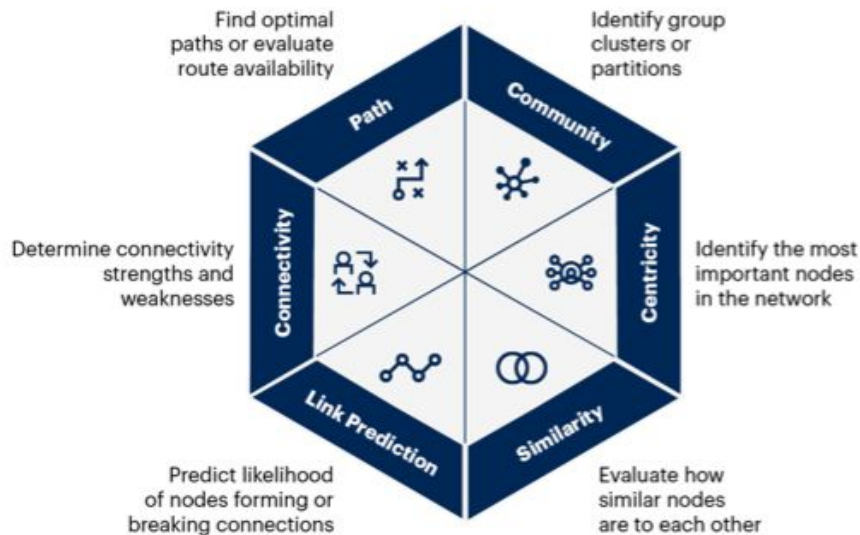
“Graph analysis is possibly the *single most effective competitive differentiator* for organizations pursuing data-driven operations and decisions after the design of data capture.”

Gartner®



Six Types of Graph Analytics

Graph can be used to analyze all sorts of relationships across all kinds of systems even beyond process or beyond the confines of individual operational models.

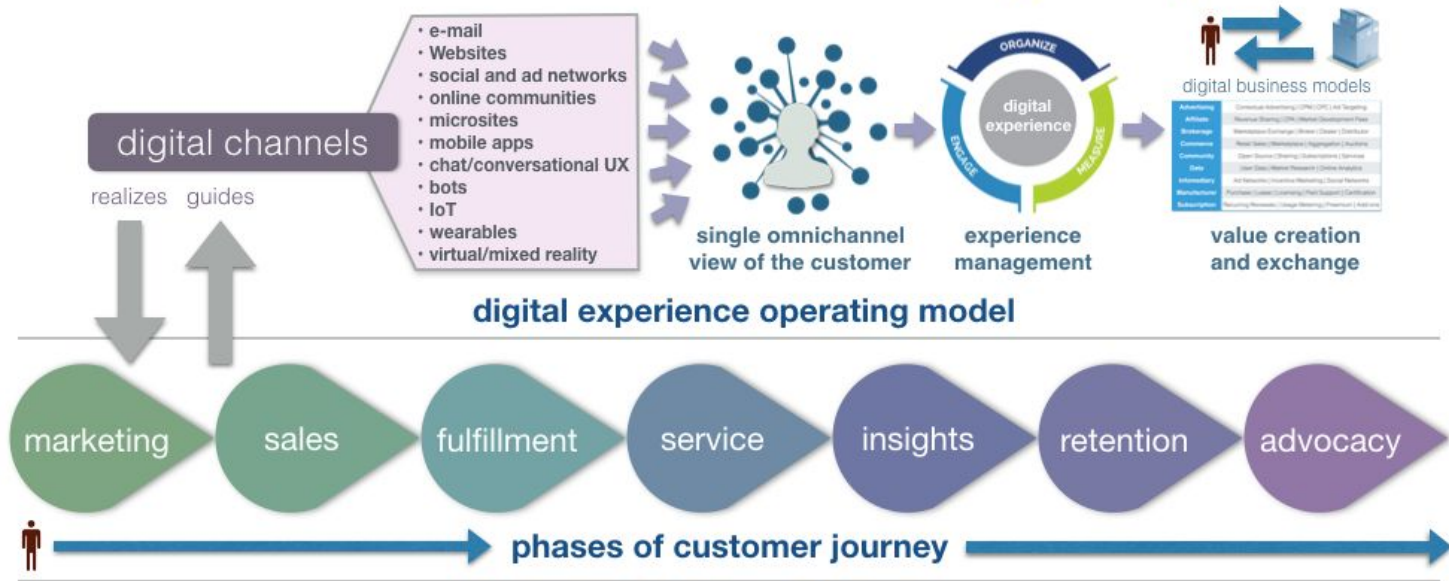


Source: "Understanding When Graph Technologies Are Best for Your Business Use Case", Jim Hare et al, 2020

Digital transformation begins and ends with the customer

“Customer experience (CX) is the product of an interaction between an organization and a customer over the duration of their relationship. This interaction is made up of three parts; the customer journey, the brand interactions, and the channels the customer uses to discover an organization's products and services, and purchase behaviors.”

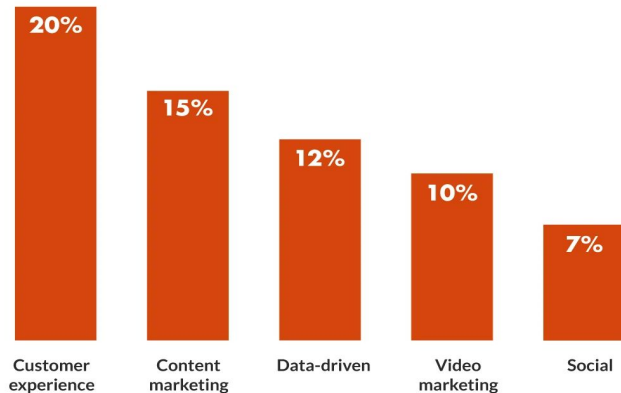
Next-Generation Customer Experiences



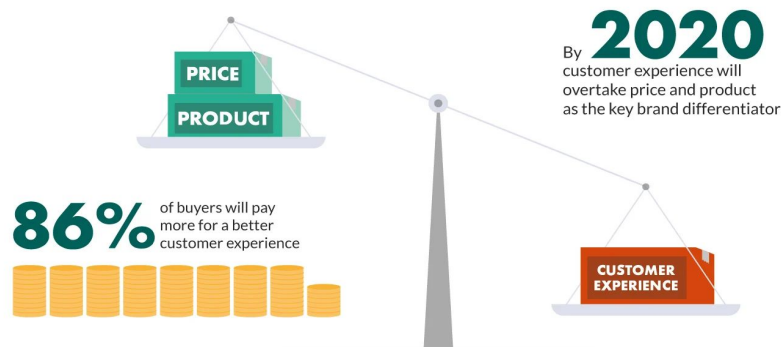
From <http://zdnet.com/blog/hinchcliffe> on by Dion Hinchcliffe

Customer Experience Trends

MOST EXCITING BUSINESS OPPORTUNITY

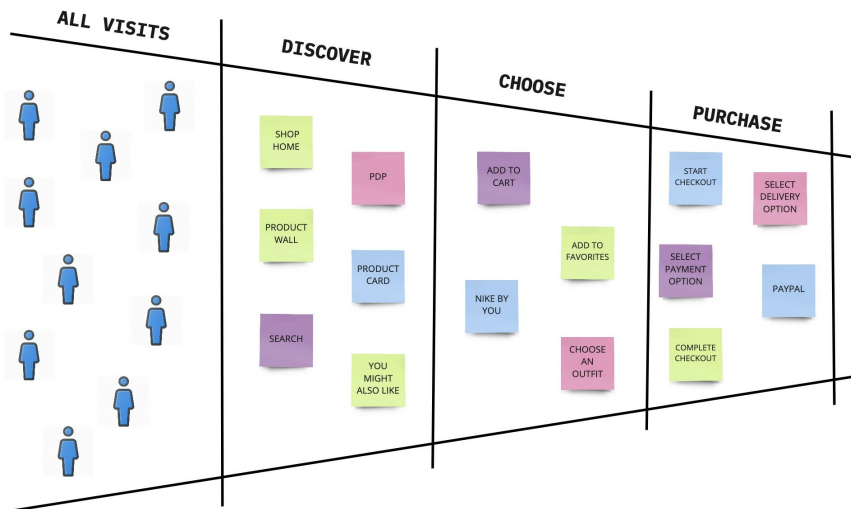


SURVEY SAYS: CUSTOMERS HIGHLY VALUE GREAT CUSTOMER EXPERIENCES



It's All About the Journey

- How do we identify the shopping stage for a person and/or groups
- What are the interaction patterns for shoppers during the “Discovery” phase?
- What is the Time-to-Purchase from first visit to purchase?
- What behaviors/patterns cause shoppers to abandon the journey?
- What Marketing activities influence moving from “Discover” to “Choose”?
- What Channels drive the most purchases?
- What are the customer Outbound preferences?



Customer Journey



Customer Experience Demonstration

Use Cases

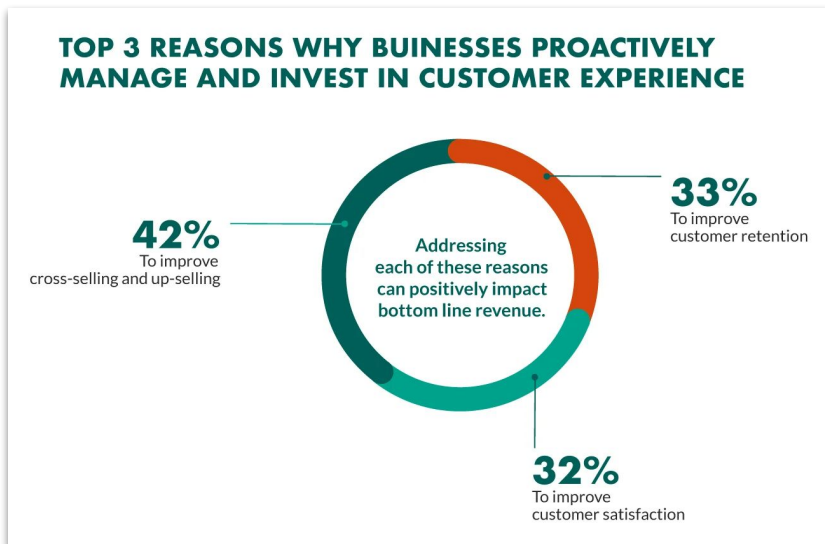
- Customer 360
- Marketing analytics
- Find best customers
- Segmentation / similar customers
- HTAP platform
- Customer journey

Hybrid Transactional and Analytical Platform

- Upsert data in real-time
- ACID-Compliant
- Post processing
- Immediately available

```
In [8]: 1 #####
2 # Address
3 #####
4 def gen_address(profile_id,source_id,prfl_date):
5
6     idx = random.randrange(1,13655)
7     geogdata = df_geo.iloc[idx]
8     address_id = int(geogdata[0])
9     address_type = faker.random_element(elements=('Home', 'Government', 'School', 'Condo', 'Home', 'Apartment', 'Bu
10     address_line = faker.street_address()
11     zipcode = str(geogdata[1])
12     city = geogdata[2]
13     state = geogdata[4]
14     lat = geogdata[9]
15     lng = geogdata[10]
16     market = geogdata[13]
17     region = geogdata[15]
18     population = int(geogdata[11])
19     store_id = int(geogdata[16])
20
21     ## Output Data to CSV
22     prfl_addr = str(profile_id) + "," + str(source_id) + "," + str(prfl_date) + "," + str(address_id) + "," + str(a
23     prfl_addr.write(prfl_addr)
24
25     ## Fill DataFrames
26     df_address.loc[profile_id] = [profile_id, source_id, prfl_date, address_id, address_type, address_line, zipcod
27     df_location.loc[profile_id] = [zipcode, lat, lng, market, region, population, store_id]
28
29     ## Upsert to GraphDB
30     graph.upsertVertex(vertexType='Address', vertexId=address_id,attributes={'address_type':address_type,'address_l
31     graph.upsertEdge(sourceVertexType='Profile', sourceVertexId=profile_id,edgeType='profile_address', targetVertex
32     graph.upsertVertex(vertexType='Location', vertexId=zipcode,
33         attributes={'lat':lat,'lng':lng,'market':market,'region':region,'population':population,
34             'store':store_id})
35     graph.upsertEdge(sourceVertexType='Address', sourceVertexId=address_id,edgeType='address_location', targetVerte
36     graph.upsertEdge(sourceVertexType='Address', sourceVertexId=address_id,edgeType='address_source', targetVertexI
37
38     return (address_id)
```

Business Impact



- **Improve Cross-Sell and Up-Sell**
 - Recommendations
 - Next best action
- **Improve Customer Satisfaction**
 - Personalization
 - Channel preferences
- **Customer Retention**
 - Purchase patterns
 - Improved offers
 - Life-time value

TigerGraph for Customer Experience

Customer Insights

Analytics to understand customer behaviors, patterns, and preferences

2

Real-Time Decisioning

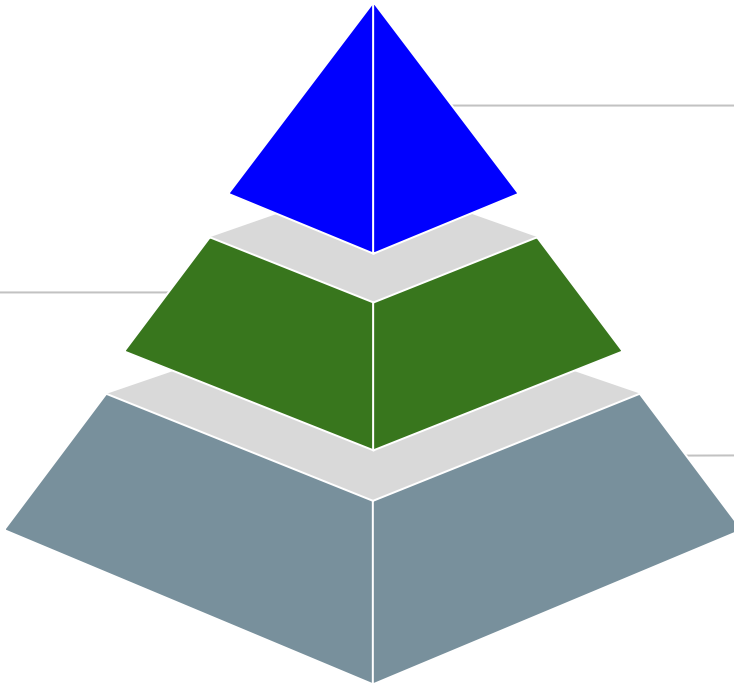
Instance decisions for product recommendations, next best action, and follow-up procedures

1

Relationship Data Management

Understand customer segments and groups to help improve customer retention and loyalty management

3



Business Impact

TigerGraph Accelerators for customer analytics and experience, recommendation use cases, and graph database (sample use cases):

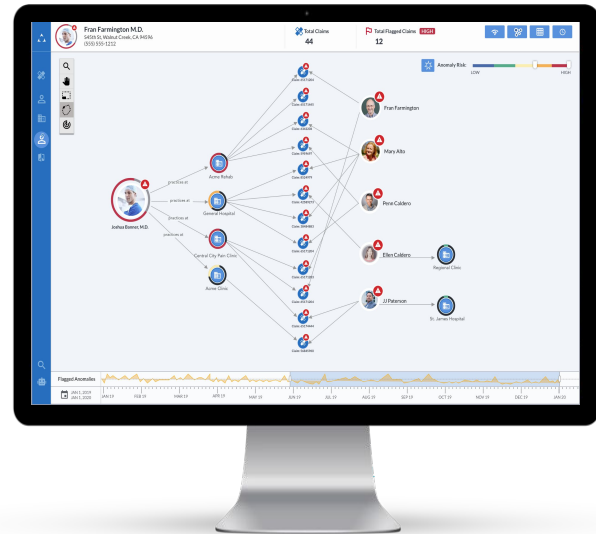
- Customer360
- Marketing Analytics
- Segmentation and Targeting
- Product Recommendations
- Customer Services Preferences
- Community Detection

Includes widgets / visualization / apps for:

- Customer Journey Visualization
- OmniChannel Analytics
- LifeTime Value

Contains foundational components for:

- Native Graph Database and Schemas
- Data Ingestion accelerators for Spark and Kafka
- Graph Algorithms for Machine Learning



Thank You



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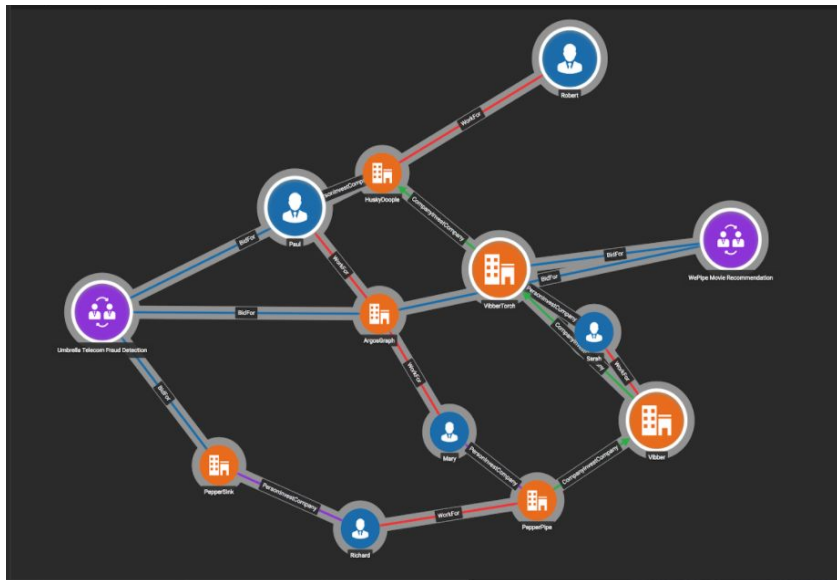


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Thank You

A Sample Graph



Graph databases consist of **vertices** and **edges**

- **Vertices** - data entities
 - for example - person, account, transaction
- **Edges** - the relationships between those entities
 - for example - person opens account, money moves from one account to another account

A graph stores the relationships between data entities - or can be used to uncover relationships between data entities