

INTRO

- Computer Engineering at Georgia Tech
- Product management at Intel (datacenter, mobile)
- Kellogg MBA
- Built a social insights and analytics company now a part of Material
- Led many client engagements
 - 1st party & social data developing customer / market insights

intel

MATERIAL+











brook@4brook.com

TA: AZIZA

- Bachelors in Finance and Data Science at MTSU
- Currently pursuing Masters in Artificial Intelligence at Northwestern
- Keenly involved in conducting research where she was part of:
 - Business and Economic Research Center at MTSU as a Research Assistant
 - Recipient of 2021 Station1 Frontiers Fellowship to participate in socially directed science and technology research.
 - Currently contributing to research at Tiilt Lab (Technological Innovations for Inclusive Learning & Teaching) from NLP point of view
- Developed EduAnalytics start up project aimed to automate the data analysis process for edu centers in Uzbekistan.

ROADMAP FOR OUR FIRST CLASS

- Course overview
- Twilio breakout
- Data models and Databases
- Database breakout
- 5 min break
- Lab intro
- i Lab

OUR GOALS

Building skills for manipulating and interpreting data

Understanding data intensive systems (tech, process, people)

Delivering business outcomes

Where does the data come from?

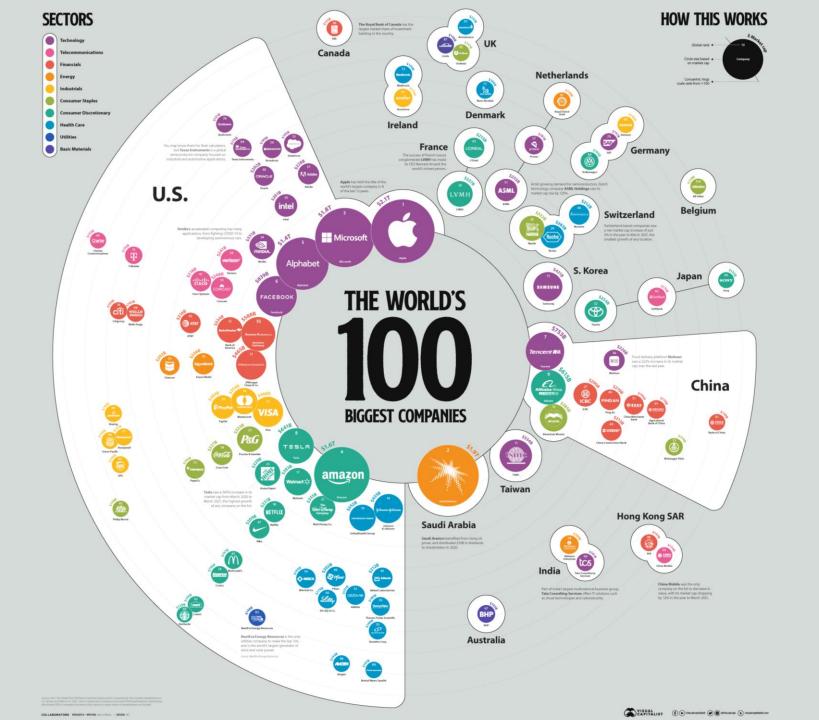
What happened to it along the way?

Does this create decision / operational / legal risks?

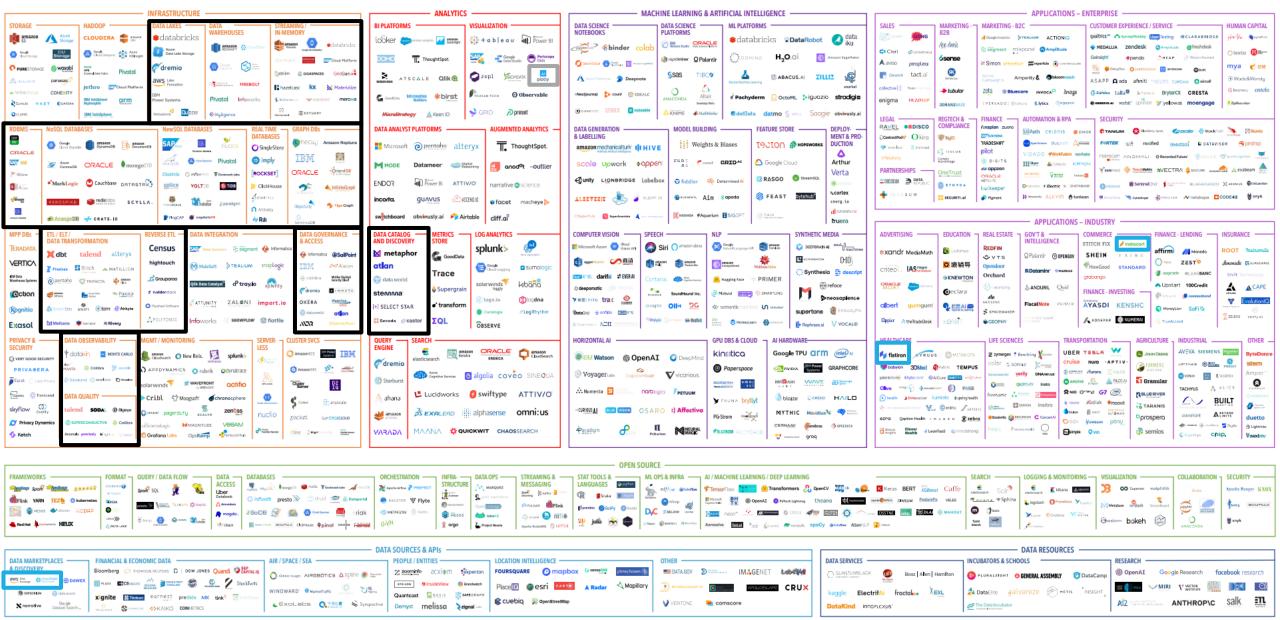


DATA INTENSIVE SYSTEMS

- Product manager / working with / leading teams building (building on top of) data intensive systems
- Working at a data intensive company
- Analyst / consumer of analysis of data

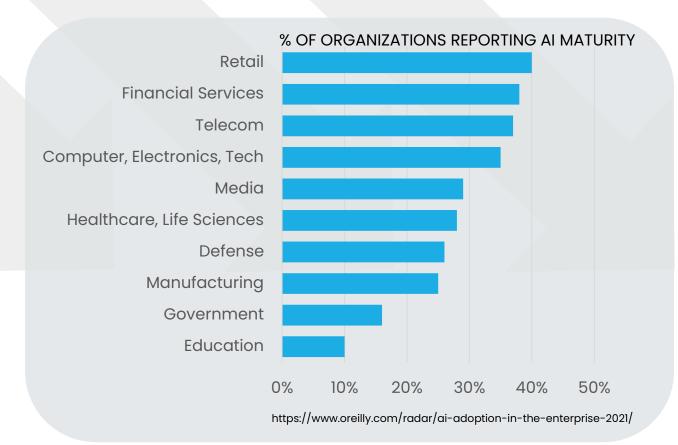


MACHINE LEARNING, ARTIFICIAL INTELLIGENCE, AND DATA (MAD) LANDSCAPE 2021



MEGA-TRENDS







INCREASED SOPHISTICATION IN HOW WE THINK ABOUT DATA









Usage	Lifecycle	Modality	Challenges
Reporting Analytics Decision Support Data Driven Products	Acquire Analyze Activate Archive	Data Model Structured Unstructured Files APIs	Access Scale (Latency,Throughput,Volume) Quality Governance Discoverability Bias



CLASS PARTICIPATION

- Grade: 10% participation, 10% attendance
 - Quality >> quantity
- Use office hours if you would like to discuss topics that seem too narrow or in-depth for class Q&A, discussion
 - We are covering a lot of ground in this class
- Avoid acronyms and jargon, like Thanos it is a bit inevitable
- Vendor agnostic, but we'll use some examples from key vendors that highlight important details
- Learning to drive != Building a car engine
 - Eventually relevant an F1 engineer will have a different perspective to the driver

REFLECTIONS

- 1 per class, 20% of grade. Groups of ~4. ~750 words +/-250, prefer link to google colab markdown doc.
 - Template
- Due by Tuesday at 10PM Central, same as homework, submit one copy via Canvas. (same as homework assignment)
- Twilio Alfredo Sone, Taylor Baker, Cameryn Steiger, Tushita Singhel
- Databases Jack Van Vleck, Hannie Gulcin, Benjamin Housley, Tiffany Dsouza
- Colab & SQL- Meet Patel, Katie Williamson, Tianyi He, Jason Feng
- Reflection will be shared to all students so you can keep copies

HOMEWORK

- 4 Paired assignments
- Available Thursday 8AM, Due Tuesday by 10pm
- Explore the data in a notebook based environment, accomplishing the key elements / techniques demonstrated in class
- Add a final markdown cell of < 250 words with your team's recommendation

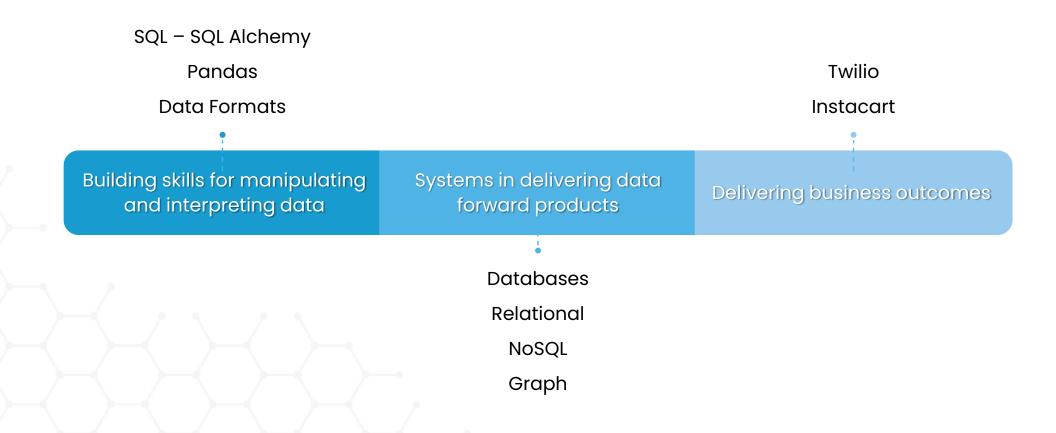
MORE INFO

- I'll post the class slides and additional readings after class
- Class liaison?
 - Please let me know the most useful time for office hours (tentatively Thursday 2PM)
 - Thursday 10:30AM central to 5:30PM central.

DATA MODELS & DATABASES

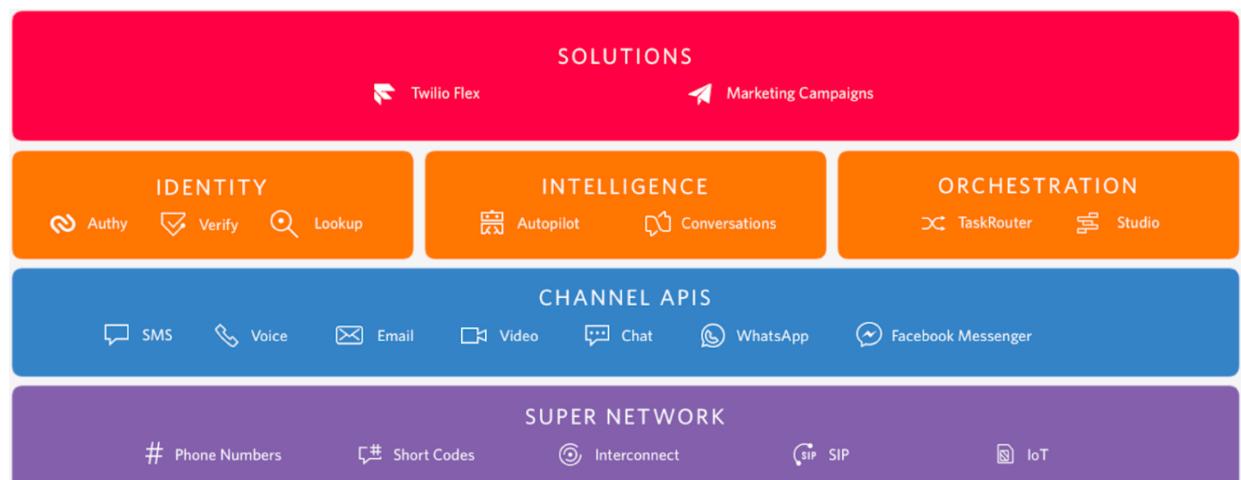


OUR GOALS - THIS CLASS



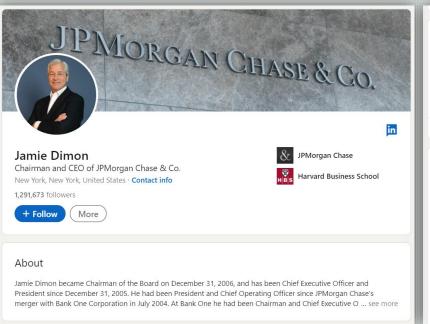


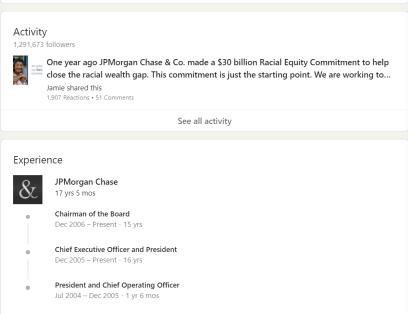


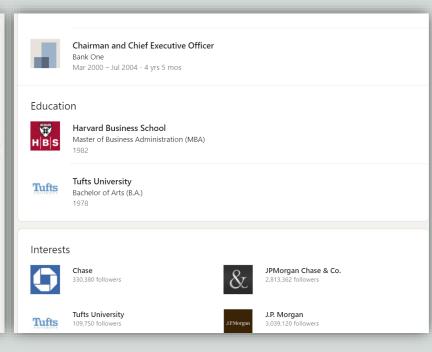


DATA MODELS

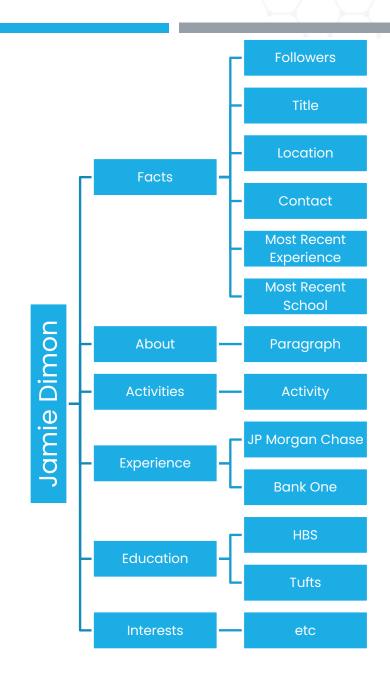
LINKEDIN PROFILE







THE DOCUMENT MODEL



THE SPREADSHEET

Rusty	Claratatatata		
	1 Rusty Sheldrick rsheldrick0@barnesandnoble.com P		Phased systemic capability
Gabi	Grenshields	ggrenshields 1@51.la	Enterprise-wide context-sensitive initiative
Godwin	Brandli	gbrandli2@infoseek.co.jp	Proactive upward-trending initiative
inette	Tease	ltease3@devhub.com	Vision-oriented reciprocal parallelism
⁄etta	Plastow	yplastow4@wired.com	Devolved zero administration intranet
/ick	Folshom	vfolshom5@youtube.com	Upgradable coherent framework
Morlee	Earlam	mearlam6@ning.com	Innovative stable website
Vilmer	Brandolini	wbrandolini7@washingtonpost.com	Secured analyzing workforce
Chaim	Van der Baaren	cvanderbaaren8@webnode.com	Realigned incremental access
Hedy	Cuttle	hcuttle9@liveinternet.ru	Managed coherent core
Alikee	O'Halloran	aohallorana@sun.com	Multi-lateral tertiary knowledge base
Melloney	Danbury	mdanburyb@people.com.cn	Extended content-based time-frame
Mallory	Hargey	mhargeyc@nyu.edu	Integrated empowering data-warehouse
Charyl	Ralling	crallingd@angelfire.com	Business-focused 5th generation middleware
Arther	Kidston	akidstone@printfriendly.com	De-engineered mission-critical implementation
Good in	odwin nette tta ck orlee ilmer naim edy ikee elloney allory	bodwin Brandli Tease tta Plastow ck Folshom orlee Earlam ilmer Brandolini van der Baaren edy Cuttle ikee O'Halloran elloney Danbury allory Hargey naryl Ralling	bodwin Brandli gbrandli2@infoseek.co.jp Tease Itease3@devhub.com tta Plastow yplastow4@wired.com ck Folshom vfolshom5@youtube.com orlee Earlam mearlam6@ning.com ilmer Brandolini wbrandolini7@washingtonpost.com raim Van der Baaren cvanderbaaren8@webnode.com edy Cuttle hcuttle9@liveinternet.ru ikee O'Halloran aohallorana@sun.com elloney Danbury mdanburyb@people.com.cn allory Hargey mhargeyc@nyu.edu naryl Ralling crallingd@angelfire.com

Data from: Mockaroo

EXTENDING THE SPREADSHEET FOR EDUCATION

Query Update

id ₹	first_name ▼	last_name ▼	email Y	education1	education2	education3
331	Scottie	Mellmer	smellmer96@icio.us	Stephens College		Gujarat Ayurved University
332	Frankie	Cramp	fcramp97@state.tx.us	Universidad Nacional de General Sarmiento	Chulalongkorn University	Rajasthan Technical University
333	Nigel	Templeman	ntempleman98@theglobeandmail.c	Ulsan University		
334	Bastien	Cersey	bcersey99@latimes.com	Ilam University	Tianjin University of Commerce	
335	Auberon	Crosham	acrosham9a@opera.com	Nkumba University		
336	Elyn	Lightfoot	elightfoot9b@cdc.gov	Technological University (Taungoo)		
337	Wilona	Borrott	wborrott9c@hp.com	Polonia University in Czêstochowa		
338	Dasie	Acum	dacum9d@soup.io	Liaoning Technical University	Medaille College	
339	Lesli	Robathon	Irobathon9e@youtube.com	National Law School of India University		
340	Frank	Chaldecott	fchaldecott9f@home.pl	Southern California Institute of Architecture		
341	lain	Roylance	iroylance9g@foxnews.com	Hanoi University of Mining and Geology	Universidad Industrial de Santander	
342	Yolane	Dorrity	ydorrity9h@sbwire.com	Perm State University		
343	Tammy	de Mendoza	tdemendoza9i@bloomberg.com	Fachhochschule Biberach, Hochschule für Bauwesen	Chunchon National University of Education	
344	Kerry	Snoday	ksnoday9j@sphinn.com	Kirikkale University	Nizhny Novgorod State Architectural - Building	
345	Robinett	Stegel	rstegel9k@baidu.com	Lagos State Polytechnic	Surugadai University	
346	Lorri	Grinham	lgrinham9l@pbs.org	Jamia Hamdard University	Life University	Universidad Peruana de Ciencias Aplicadas
347	Clem	Bartunek	cbartunek9m@tiny.cc	Florida Community College at Jacksonville	University of Minnesota - Crookston	
348	Humfried	Sinfield	hsinfield9n@cbsnews.com	Istanbul Ticaret University		
349	Jud	Posvner	jposvner9o@pen.io	Gorno-Altaisk State University	Kodolanyi Janos University College	
350	Kalina	Attock	kattock9p@npr.org	University College Dublin		
351	Burlie	Moseley	bmoseley9q@rambler.ru	Universiteit Doesburg (UNDO)	Ecole Nationale Supérieure du Pétrole et des Mo	
352	Jeffy	Balding	jbalding9r@meetup.com	Wah Medical College	Al-Buraimi University College	Hardin-Simmons University

if (student.education) == target & student.education2 == ...

THE RELATIONAL MODEL

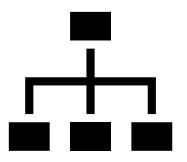
•				
id ₹	first_name ▼	last_name ▼	email	W.
331	Scottie	Mellmer	smellmer96@icio.us	
332	Frankie	Cramp	fcramp97@state.tx.us	
333	Nigel	Templeman	ntempleman98@theglobeandma	il.c
334	Bastien	Cersey	bcersey99@latimes.com	
335	Auberon	Crosham	acrosham9a@opera.com	
336	Elyn	Lightfoot	elightfoot9b@cdc.gov	
337	Wilona	Borrott	wborrott9c@hp.com	
338	Dasie	Acum	dacum9d@soup.io	
339	Lesli	Robathon	Irobathon9e@youtube.com	
340	Frank	Chaldecott	fchaldecott9f@home.pl	
341	lain	Roylance	iroylance9g@foxnews.com	
342	Yolane	Dorrity	ydorrity9h@sbwire.com	
343	Tammy	de Mendoza	tdemendoza9i@bloomberg.com	
344	Kerry	Snoday	ksnoday9j@sphinn.com	
345	Robinett	Stegel	rstegel9k@baidu.com	
346	Lorri	Grinham	lgrinham9l@pbs.org	
347	Clem	Bartunek	cbartunek9m@tiny.cc	
348	Humfried	Sinfield	hsinfield9n@cbsnews.com	
349	Jud	Posvner	jposvner9o@pen.io	
350	Kalina	Attock	kattock9p@npr.org	
351	Burlie	Moseley	bmoseley9q@rambler.ru	
352	Jeffy	Balding	jbalding9r@meetup.com	

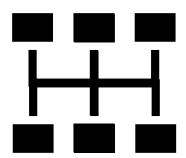
student_id	institution_id
331	45
331	12

ld ₹	Institution
1	3rd Military Medical University
2	École Nationale Supérieure d'Ingénieurs de Constructions Aéronautiques
3	Abasyn University Peshawar
4	Abubakar Tafawa Balewa University
5	Academy of Economics in Wroclaw
6	Academy of Public Administration
7	Adamson University
8	Adeyemi College of Education
9	Afghan University
10	Air University
11	Akhbar El Yom Academy
12	Al Nahrain University
13	Alabama Agricultural and Mechanical University
14	Alain University of Science and Technology
15	Al-Birony University
16	Alfaisal University
17	Alhamd Islamic University
18	All India Institute of Medical Sciences
19	American College of Thessaloniki
20	American University Extension, Okinawa
21	American University of the Caribbean

SPARESENESS -> REDUCED RESOURCES
DRY - NORMALIZED
CONSISTENT UPDATES

RELATIONSHIPS





One to many

Customer -> Orders

Many to many

People -> Institutions

What constraints need to be enforced to ensure our relationships are valid?

ACQUIRING DATA







STORING FILES CONSUMING DATABASES

MOVING APIS

STORING DATA: FILES

- Common desktop: Xlsx, google sheets, csv
- Meta data: Exif for images, parquet for data
- Compression, Row vs. column locality
- Many proprietary and performance optimized formats
 - Commonly used and supported by pandas: csv, json arrays, parquet

JSON (TWITTER TIMELINE)

```
}, {
    "entryId": "tweet-1478814463629271040",
    "sortIndex": "1478814463629271040",
   "content": {
        "entryType": "TimelineTimelineItem",
       "itemContent": {
           "itemType": "TimelineTweet",
           "tweet_results": {
               "result": {
                   "__typename": "Tweet",
                   "rest_id": "1478814463629271040",
                   "core": {
                       "user_results": {
                           "result": {
                               "__typename": "User",
                               "id": "VXNlcjo4NjYyNjg0NQ==",
                               "rest_id": "86626845",
                               "affiliates_highlighted_label": {},
                               "has_nft_avatar": false,
                               "legacy": {
                                   "blocked_by": false,
                                   "blocking": false,
                                   "can_dm": false,
                                   "can_media_tag": false,
                                   "created_at": "Sun Nov 01 00:02:55 +0000 2009",
                                   "default_profile": false,
                                   "default_profile_image": false,
                                   "description": "physician-scientist, author, editor \nhttps://t.co/vwstu2BYrC",
                                   "entities": {
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                                          "urls": [{
                                              "display_url": "drerictopol.com",
                                               "expanded_url": "http://drerictopol.com",
                                              "url": "https://t.co/vwstu2BYrC",
                                              "indices": [38, 61]
                                          }]
                                       },
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                                              "indices": [0, 23]
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                                   "fast_followers_count": 0,
                                   "favourites_count": 40634,
                                   "follow_request_sent": false,
                                   "followed_by": false,
                                   "followers_count": 593000,
                                   "following": true,
                                   "friends_count": 534,
                                   "has_custom_timelines": true,
                                   "is_translator": false,
                                   "listed_count": 11192,
                                   "location": "La Jolla, CA",
                                   "media_count": 24527,
                                   "muting": false,
                                   "name": "Eric Topol",
                                   "normal followers count": 593000,
```

RELATIONAL DATABASES

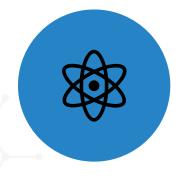
- Schema of tables and relations
- SCHEMA: This describes how real-world entities are modeled in the database. Tables, relations, and constraints (what data will we not allow in)
- Query with SQL
- Most support: text, JSON, binary data types / columns
- Optimized for storage, compute & query



Source: https://mattturck.com/data2021/



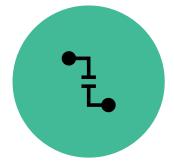
ACID TRANSACTIONS







CONSISTENCY



ISOLATION



DURABILITY

DOCUMENT ORIENTED DATABASES (TYPICALLY NoSQL)

- Data typically stored in json / document model
- API: simplest query using key / id fetch typically support indexing and partitioning across pre-defined keys
- Many now support sql



Source: https://mattturck.com/data2021/



GRAPH DBs



Amazon Neptune













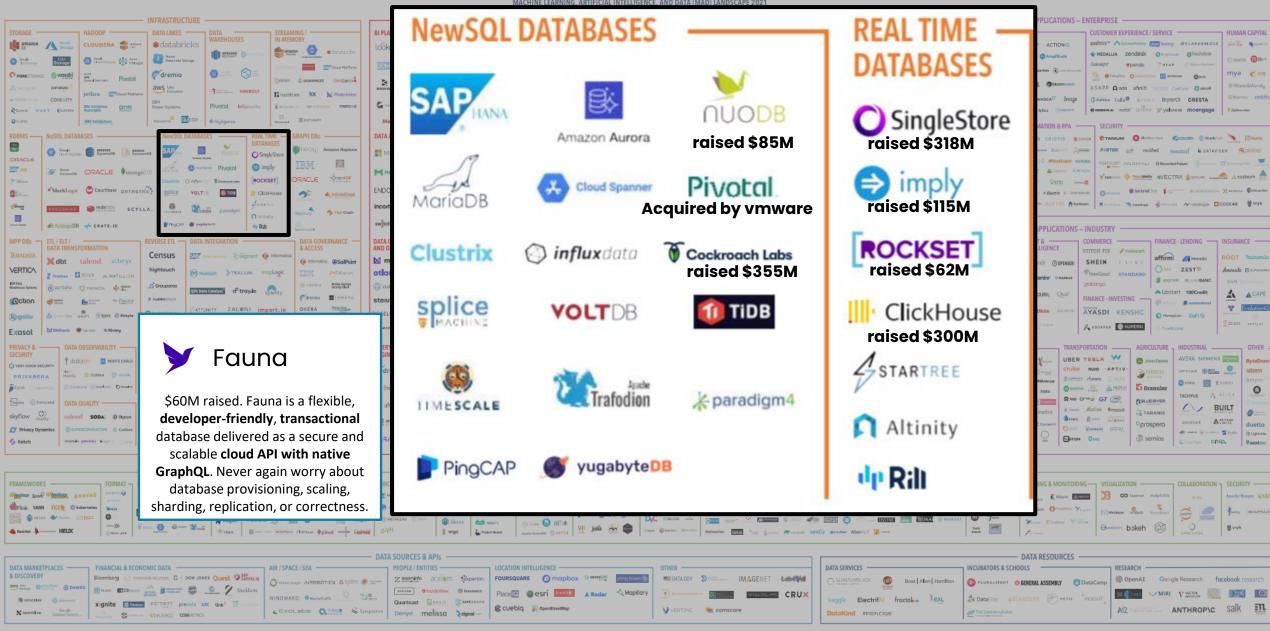






GRAPH DATABASES

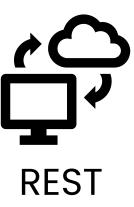
- A graph database stores nodes and relationships instead of tables, or documents.
- Nodes (data) and edges (relationships)
- Data is stored without restricting it to a pre-defined model, perhaps multiple data models
- Facebook Tao





APIS

- Databases typically support proprietary data connectors that are language / library dependent
 - ODBC/JDBC
- REST JSON using http verbs
 - CRUD
- Protobuf
- GraphQL







BREAK





ML strategy tip

When you have a problem, build two solutions - a deep Bayesian transformer running on multicloud Kubernetes and a SQL query built on a stack of egregiously oversimplifying assumptions. Put one on your resume, the other in production. Everyone goes home happy.

3:45 AM · Aug 12, 2021 · Twitter for iPhone

696 Retweets **114** Quote Tweets **4,081** Likes

• • •

orders

order_id	user_id	eval_set	order_number	order_dow	order_hour_of_day	days_since_prior_order
Y	Y	Y	∇	Y	7	Y
2398795	1	prior	2	3	7	15
2254736	1	prior	4	4	7	29
3367565	1	prior	6	2	7	19
3108588	1	prior	8	1	14	14
2550362	1	prior	10	4	8	30
2168274	2	prior	1	2	11	nan
1901567	2	prior	3	1	10	3
1673511	2	prior	5	3	11	8
3194192	2	prior	7	2	12	14
1718559	2	prior	9	2	9	8
1402090	2	nrior	11	1	10	30

aisles

aisle_id	aisle
Y	Y
1	prepared soups salads
2	specialty cheeses
3	energy granola bars
4	instant foods
5	marinades meat prepar
6	other
7	packaged meat
8	bakery desserts
9	pasta sauce
10	kitchen supplies

primary key

foreign key

	order_id	product_id	add_to_cart_order	reordered
	7	∇	Y	Y
У	2	33120	1	1
	2	9327	3	0
	2	30035	5	0
	2	40141	7	1
	2	43668	9	0
	3	24838	2	1
	3	21903	4	1
	3	46667	6	1
	3	32665	8	1
	4	26434	2	1
	4	27761	4	1

product_id	product_name	aisle_id	department_id
Y	7	Y	Y
2	All-Seasons Salt	104	13
4	Smart Ones Classic Favor	38	1
6	Dry Nose Oil	11	11
8	Cut Russet Potatoes Stea	116	1
10	Sparkling Orange Juice &	115	7
12	Chocolate Fudge Layer Ca	119	1
14	Fresh Scent Dishwasher C	74	17
16	Mint Chocolate Flavored	103	19
18	Pizza for One Suprema Fr	79	1
20	Pomegranate Cranberry &	98	7
22	Fresh Breath Oral Rinse	20	11

department_id	department
Y	Y
1	frozen
2	other
3	bakery
4	produce
5	alcohol
6	international
7	beverages
8	pets
9	dry goods pas
10	hu1k

products

departments