

Stanford CS 520 Knowledge Graphs

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(allegrograph.com)



About Franz Inc.

- Privately held, Self-funded, Profitable
- Headquartered: **Oakland, CA**
- Flagship product: **AllegroGraph**
- Compiler History: **Common Lisp and Prolog**
for Artificial Intelligence and Complexity
- Industries: **Healthcare, Intelligence/Defense, Pharma/Biotech**
- Key Customers: **Montefiore, BAE, 2Is, BlueCross/BlueShield, Novartis, Credit Suisse, AstraZeneca**
- Strategic Partners: **BAE, Intel, Cloudera, Montefiore Health Systems Pool Party, 2Is, Smart Logic, TopQuadrant, Expert Systems**



For today – just a simple introduction

- What is in a Knowledge Graph
- Use case we are involved with
 - Document and Text based Knowledge Graphs
 - Event Based Knowledge Graphs.
- Demo of
 - Chomsky Knowledge Graph
 - Large Hospital Use Case

What is in a modern knowledge graph?

- A semantic graph database, scalable, secure, acid
- **Ontologies and taxonomies**
- Reasoning and rule based processing
- Smart integration of silos of information
- Machine Learning and Advanced Analytics
- Natural Language Processing and Text classification

And recently: more and more:

- Speech recognition
- Chatbots.

We use our KG platform in various domains

- Do you want to live forever?
 - Biography Knowledge Graphs for interesting (and/or public) figures



- Predict risk in the automotive supply chain



We have solutions in various domains

- What do my agents talk about? What conversation style leads to more sales?
 - A knowledge graph for Intelligent Call Centers
- How can we make our data integration, our ETL, queries, feature extraction and holistic overview a 100 times easier?
 - An events based knowledge graph.



Addressing value-based healthcare with Intel® Xeon® processors and Franz AllegroGraph*

Challenge

Located in the Bronx, Montefiore Health System serves one of the most ethnically and socioeconomically diverse populations in the US. The complex includes the Montefiore Medical Center, the Albert Einstein College of Medicine, and a research facility. Unlike a pay-per-service model, as an accountable care organization Montefiore delivers value based on patients' long-term health—during their hospital or clinic visit and after they return to the community.



Two main types of KG

- Document and NLP based knowledge graphs
 - Chomsky Knowledge Graph
- Entity/Event based knowledge graphs.
 - Montefiore

The Chomsky Knowledge Graph

Overall

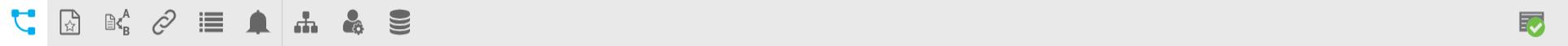
- Preserve Chomsky's legacy

Practical

- Make it easy for journalists, students, researchers, strategists and politicians to find 'his thoughts'

Our platform solution:

- Knows everything about what Chomsky ever said or wrote, what others wrote about him, how concepts he talked about connect 'internally' and 'externally' (that is: to other knowledge sources, including linked open data, news paper articles)



- Chomsky Project
 - Concepts (3)
 - Equality/Inequality (3)
 - Freedoms (2)
 - power (2)
 - Hard Power (4)
 - Attacks on Individuals (6)
 - Military Attack (4)
 - Violent Force (0)
 - War (2)
 - Cold War Proxy Wars (3)
 - Unwinnable War (4)
 - Soft Power (5)
 - Events (4)
 - Legislation (3)
 - Media (4)
 - Nation States/Geographic Areas (23)
 - Organizations (8)
 - People (4)
 - Lists
 - Collections
 - GraphEditors

Cold War Proxy Wars

<https://allegrograph.poolparty.biz/ChomskyProject/525>

- + Add to Collection
- ⊖ Add to Blacklist
- ⊖ Add to ExactMatch
- ⊗ Delete Concept

Details Notes Documents Linked Data Triples Visualization Quality Management History

SKOS ⓘ +

Broader Concepts

[War](#) ⓘ

Narrower Concepts

- ⊗ [Korean War](#)
- ⊗ [Nicaraguan Revolution](#)
- ⊗ [Vietnam War](#)

ⓘ ⓘ +

Related Concepts

- ⊗ [Soviet Containment](#)

ⓘ

Top Concept of Concept Schemes

ⓘ

Preferred Label

⊖ Cold War Proxy Wars en

Alternative Labels

⊕

Hidden Labels

⊕

Scope Notes

⊕

Definitions

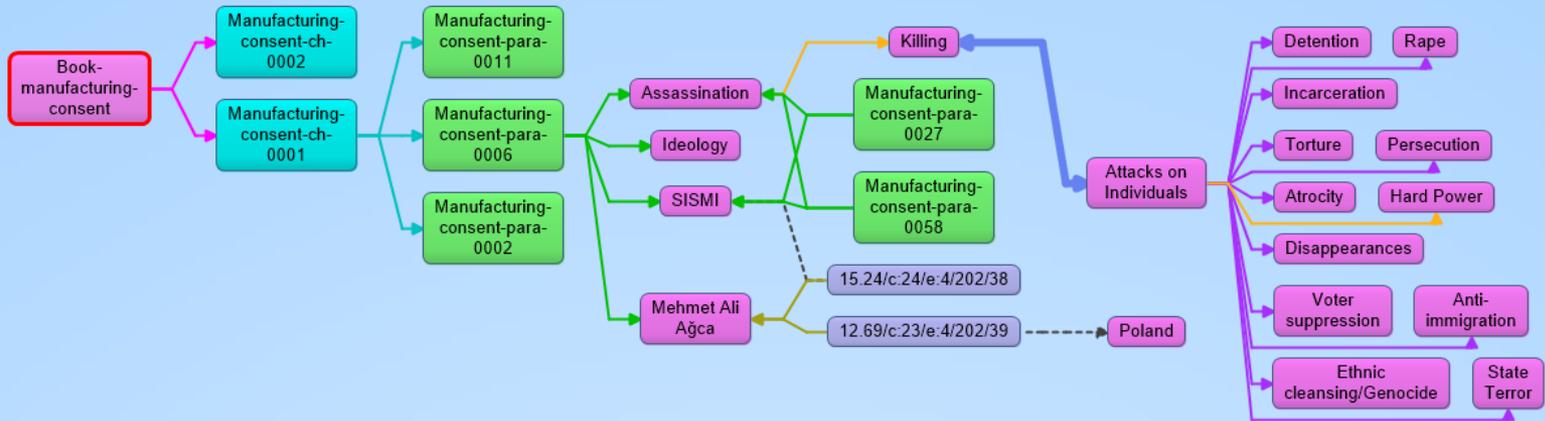
⊕

A glance at the knowledge graph

Gruff 7.2.1 on AllegroGraph 6.4.1 (server 6.4.2) internal-testing:chomsky read-only 23,843 triples server flux.franz.com
File View Text Search Display Link Remove Layout Select Edit Global Options Visual Graph Options Help

Broader
From
Has Chapter
Has Concept
Has Paragraph
Narrower
To
Multiple Predicates

Book
Chapter
Concept
Odds Ratio
Paragraph



An events based approach
to integrate
enterprise data

An event based knowledge graph in healthcare

SOLUTION BRIEF

Healthcare and Life Sciences
Data Analytics Solutions



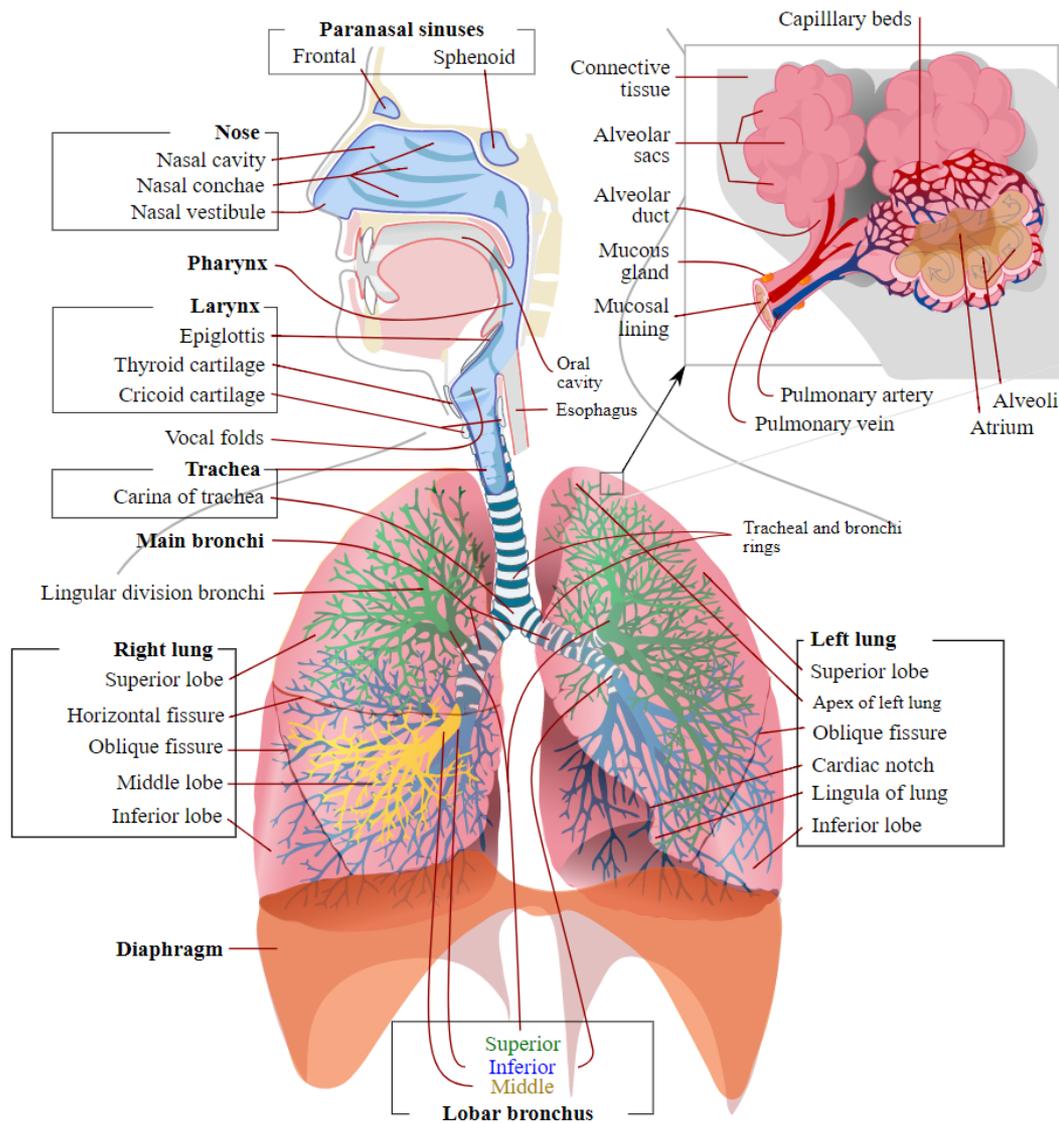
Montefiore Creates Data Analytics Platform to Advance Patient Care

Addressing value-based healthcare with Intel® Xeon® processors and Franz AllegroGraph*

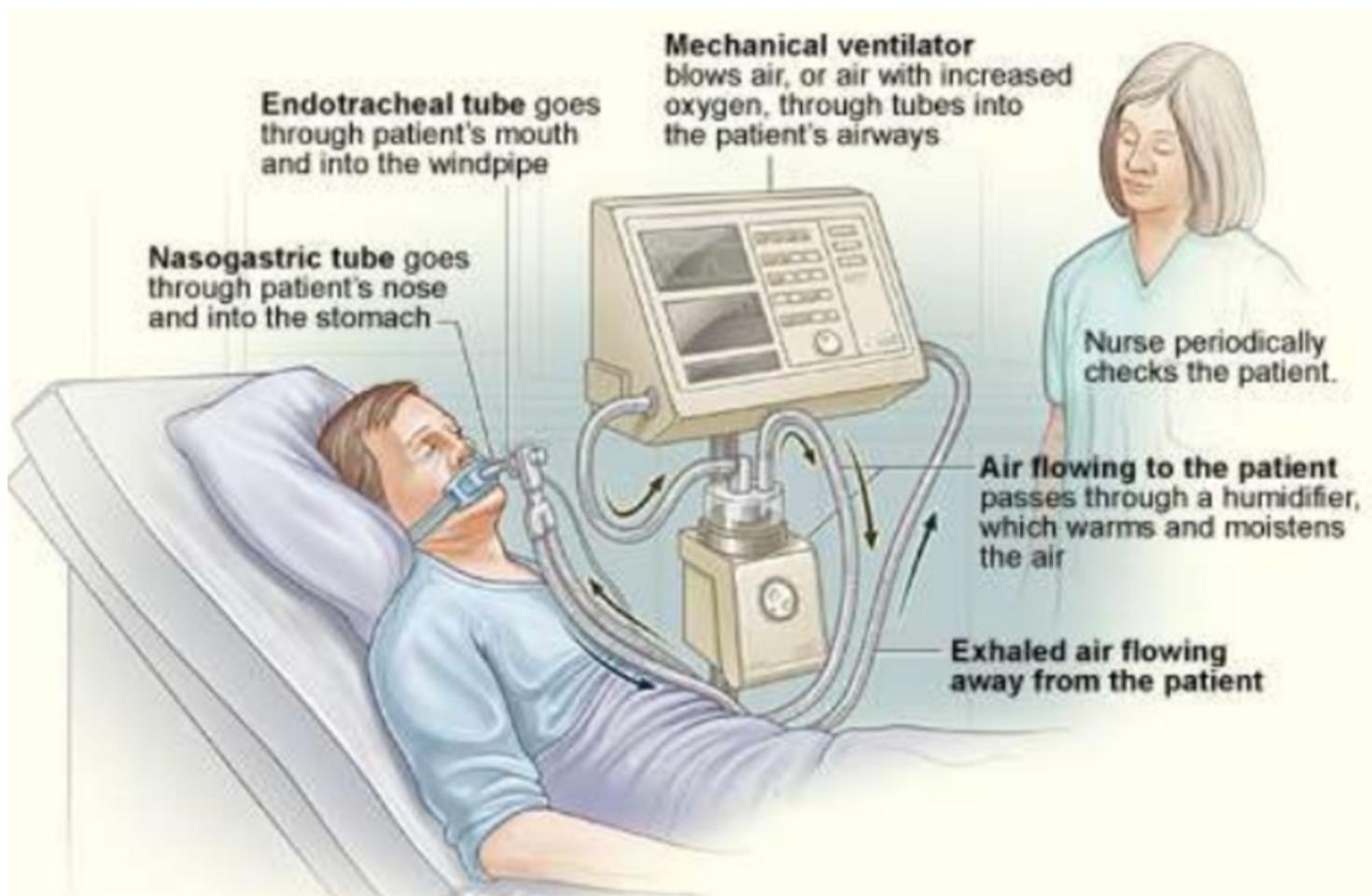
Challenge

Located in the Bronx, Montefiore Health System serves one of the most ethnically and socioeconomically diverse populations in the US. The complex includes the Montefiore Medical Center, the Albert Einstein College of Medicine, and a research facility. Unlike a pay-per-service model, as an accountable care organization Montefiore delivers value based on patients' long-term health—during their hospital or clinic visit and after they return to the community.

Respiratory failure is very dangerous. More than 40 % mortality even if it occurs in hospital

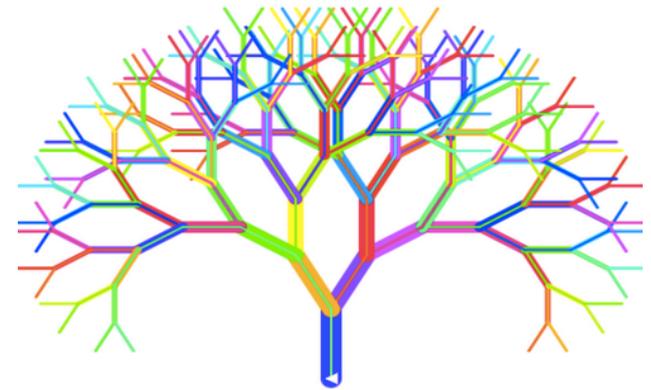


Needs instant intubation if at risk; but when?



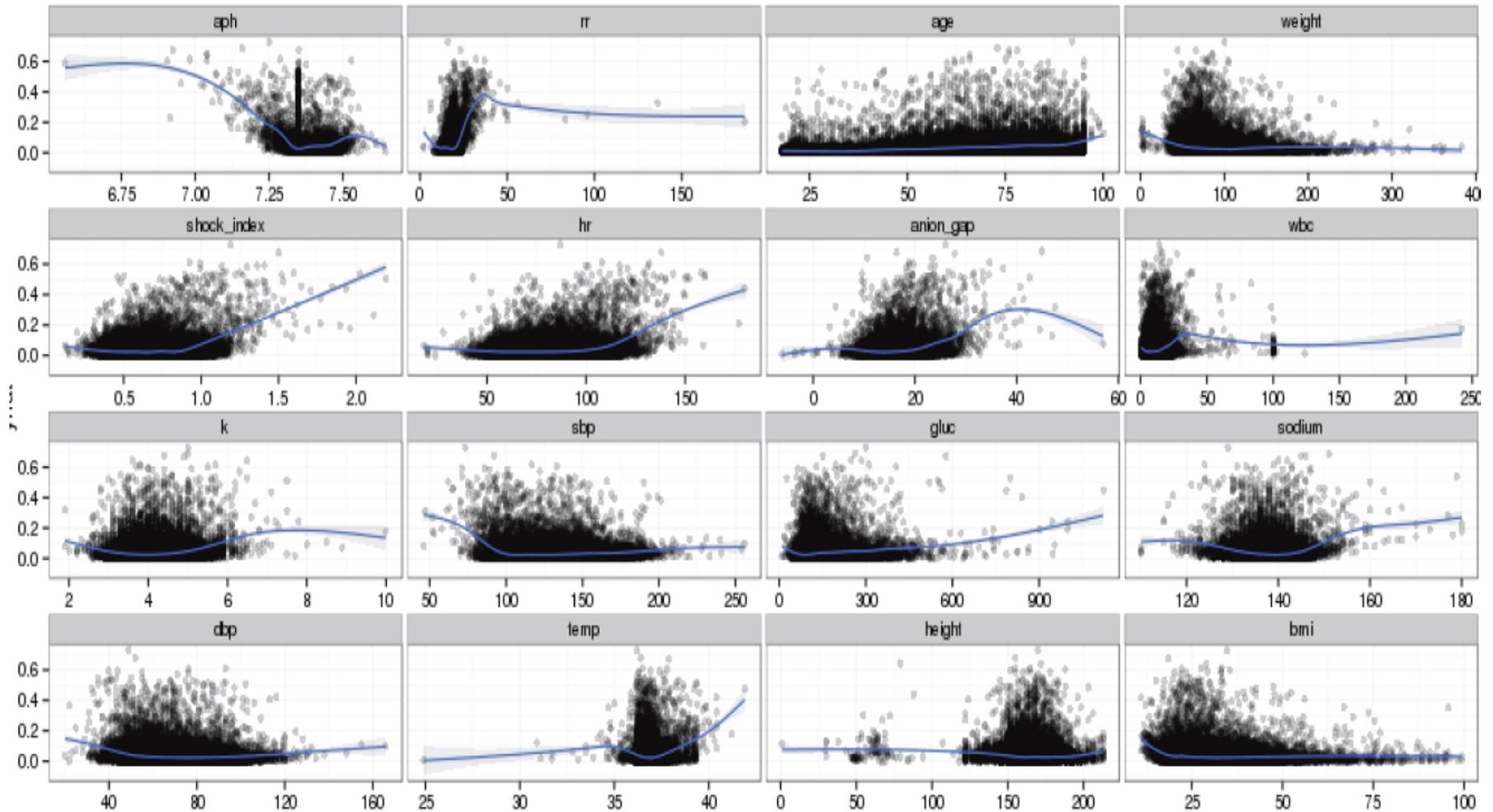
First production use case: a predictive model for respiratory failure on top our Knowledge Graph

- A random forest model detects respiratory failure up to 48 hours before the event.
Faster than doctors and nurses
- The model uses 46 complex variables, a doctor in general looks at 3 to 4 variables.
 - A mix of lab results, (real time) measurements, diagnostics, other
 - Complex: give me largest difference between the value for SerumCalciumLevel and the midpoint 9.5 in the range of 0 to 20 for the last 24 hours.
- Greatly reduce unnecessary intubation with same clinical outcome
- Saves many millions per year



Case Study
1

This is why it is hard for a clinician: every variable has its Distribution and relationship to 'risk'



▼ Evaluate need for critical care consult — **Required**

▼ Critical care consult or contraindication — **Required**

Indications for critical care consult:

- Patient identified as at increased risk for acute respiratory failure or death in the hospital by APPROVE
- Patient is receiving or being evaluated for Non-Invasive Ventilation (NIV) for respiratory dysfunction
- Any concern for patient
- Even if previously seen by critical care, Critical Care consult can be called back as needed.

Can also page overhead for Rapid Response for emergencies

- Inpatient consult to critical care
- Reason for no critical care consult

Report: FS

Summary

Problem List

Index

- Current Meds
- I&O and Results
- All Labs - Last 72 hr
- All Imaging - Last 72 hr
- Diet Orders
- Active Lines
- Problem List
- Allergies
- History
- Treatment Team
- SmartLinks and SmartPhrases
- ICD-10 Diagnosis
- SmartPhrases
- My Charges & Notes
- All Notes - Last 72 hr
- EMG SmartLinks and Provider Checklist
- SmartPhrases

24 Hrs	8 Hrs	4 Hrs	1 Hr	15 Min	All
701 - 01/20 0700	1501	1901			
6.1)		98.2 (36...	Temp		
illary		Oral	Temp...		
99		97	Heart...		
onitor			Heart...		
19		18	Resp r...		
167		06/17	RR (w/...		

▼ Evaluate for potential sepsis or infection — **Required**

▼ Severe sepsis/septic shock orders or contraindication — **Required**

- Will Consider severe sepsis/septic shock order set
- Reason for no sepsis order set
- Severe sepsis/septic shock order set already activated

▼ Consider goals of care discussion — **Required**

▼ Goals of care discussion — **Required**

- Goals of care discussion took place with patient and/or family in the past 48 hours
- Goals of care discussion will take place in the next 24 hours
- Reason for not discussing goals of care

▼ Evaluate need for palliative care consult — **Required**

▼ Palliative care consult — **Required**

Are any of the following true?

- Patient has pre-existing condition which may lead to death in the next 6 months.
- Patient has difficult to control physical or psychological symptoms like pain, shortness of breath, etc.
- Patient's goals of care are unclear.
- There is disagreement between patient, family, and hospital staff concerning major treatment decision or preference for resuscitation or life-support.

If any of the above are true, patient may benefit from palliative care services.

- Inpatient consult to palliative care
- Reason for not ordering palliative care consult

Daily Care

Transfer

Discharge

Procedure

FYI

More Activities

MICHELLE G.

Ask me anything

Links

1/20/2017

Prev APPROVESCORE=0.26 on 12/27/12

Prev APPROVESCORE=0.24 on 12/27/12

Acknowledge reason:

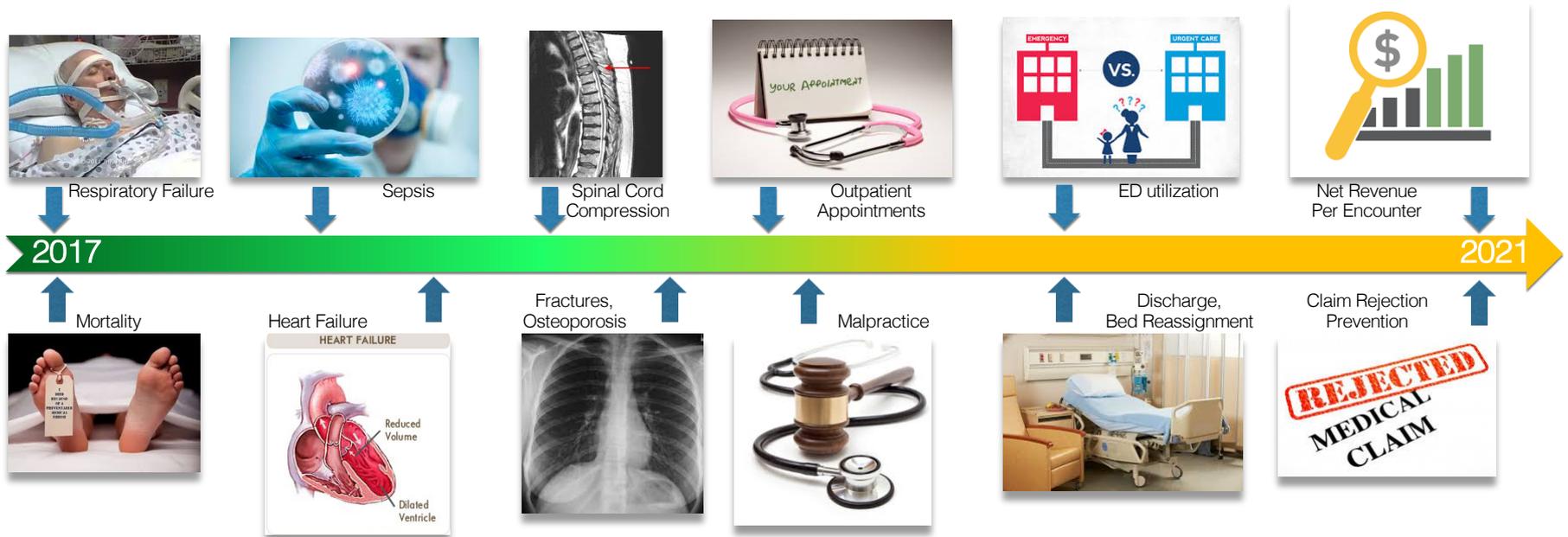
Not primary prov

Pt not changing;

Open Order Set: At-risk for acute respi

Knowledge Graph Roadmap in Montefiore

- Developed using grant funding (NHLBI, PCORI, and ICTR) and Intel/Franz collaboration
- Fully integrated with Epic, and hosted by MIT Data Center.
- Go-Live January 2017 with Respiratory Failure and Mortality Prediction → Prevention
- C: Sepsis, HF, Spinal Cord Compression (etc) all with **associated ROI**
- Wide and multi-disciplinary spectrum of applications confirms a **platform** approach



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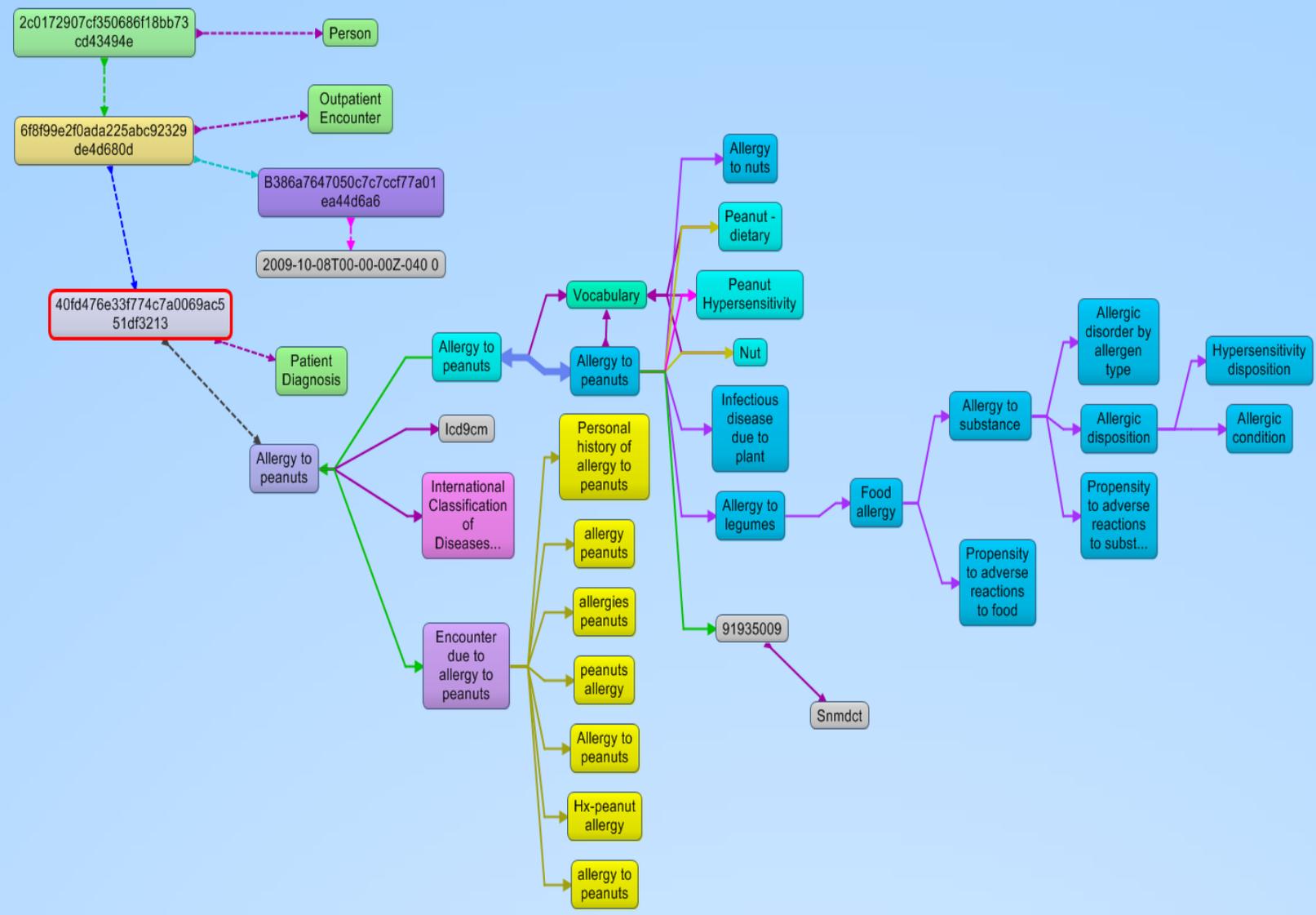
Events directly connect to Entity

The KG provides radical simplification of the EDW schema:

we turn everything into an EVENT

- Healthcare: everything that can happen to a patient is a time based (sub) event: Check In, Check Out, Test, Diagnosis, Procedure, Medication administration, Medication order, Sensor reading for vital signs, Invoice, Bill payment, Non-bill payment, all insurance interactions.
- Telecom: everything that happens with a telco user is a time based (sub) event: telephone call, sms, whatsapp, web site visit, location record, crm call, bill pay, non-bill pay ...
- Even your demographic features are events (names, address, etc)
- **From thousands of tables to one event table (well, graph)**

- Alt Label
- Begins
- Corresponds To
- Diagnosis
- Encounter
- Exact Match
- Exact Match Of
- Has Causative Agent
- In Calendar Clock
- In Scheme
- Is A
- Sy
- Type
- Multiple Predicates
- Class
- Concept
- Concept Scheme
- Finding
- OMOP Concept
- Outpatient Encounter
- Patient Diagnosis
- Patient Encounter Seen Time
- Person
- Standard Concept
- UMLS Source Abbreviation(SA)

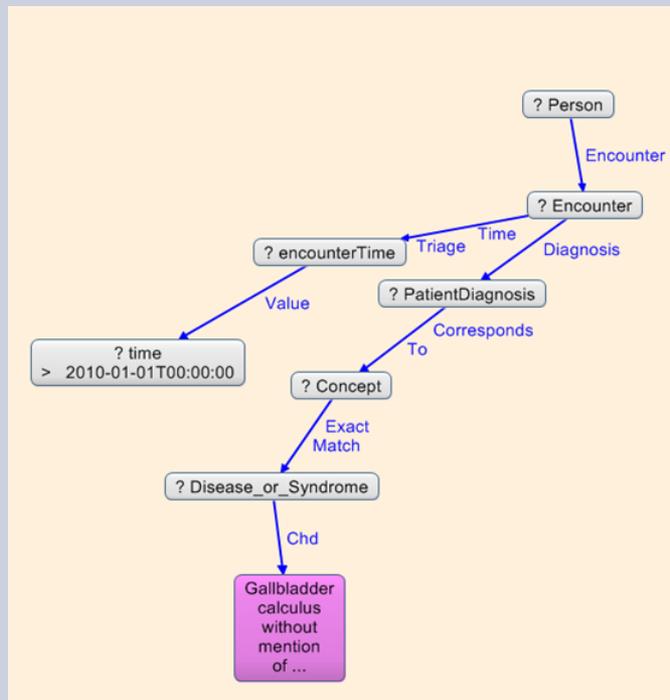


Queries are dramatically simplified

How many patients had 'gallbladder calculus' in 2010 and later

SPARQL

```
Select (count(distinct ?pt) as ?count) where {  
  ?pt cdm:encounter ?enc .  
  ?enc timew3c:begins/upper:value ?date ;  
  cdm:diagnosis/upper:correspondsTo/skos:exactMatch ?cui.  
  filter (?date > '2010-01-01T00:00:00'^^xsd:dateTime)  
  ?cui mth:chd* mth:C0497327.}
```



SQL

(Part 1 of 3)

```
Select dx.dxCode, dx.dxCount, cl.dxClassName, cl.ClassCount  
From (select distinct c.person_source_value as person,  
  c.condition_source_concept_id as dxCode, f.concept_name as  
  dxName,  
  a.concept_id_2 as dxClass, e.concept_name as dxClassName from  
  omopv5.concept_relationship a  
  join cdrn.condition_occurrence c on  
  a.concept_id_1=c.condition_source_concept_id join  
  omopv5.concept_ancestor d  
  on d.descendant_concept_id=a.concept_id_2 join omopv5.concept e  
  on e.concept_id=d.ancestor_concept_id  
  join omopv5.concept f on f.concept_id=d.descendant_concept_id  
  where a.relationship_id='Maps to' and  
  d.min_levels_of_separation=2 and e.standard_concept='S') dx JOIN  
(select distinct c.person_source_value as person,  
  c.condition_source_concept_id as dxCode, f.concept_name as  
  dxClassName from omopv5.concept_relationship a join  
  cdrn.condition_occurrence c on  
  a.concept_id_1=c.condition_source_concept_id join  
  omopv5.concept_ancestor d on  
  d.descendant_concept_id=a.concept_id_2  
  join omopv5.concept e on e.concept_id=d.ancestor_concept_id join  
  omopv5.concept f on f.concept_id=d.descendant_concept_id  
  where a.relationship_id='Maps to' and d.min_levels_of_separation=2  
  and e.standard_concept='S') cl on dx.dxClass=cl.dxClass  
order by cl.dxClass desc
```

And accessible as a key/value store for a 360 view of a patient

- It takes a one liner to get all data and knowledge for one patient in < 1 s

The screenshot shows the AllegroGraph 6.2.1 interface. The title bar indicates the server is 'small-1-8' with 592,155,683 triples. The interface includes a menu bar (File, View, Text Search, Display, Edit, Global Options, Query Options, Table Options, Help) and a toolbar with buttons for 'Run Query', 'Reindent', 'Name Query', 'Revisit', 'Graph View', 'Table View', and 'Graphical Query View'. The query input field contains the following SPARQL query:

```
select * { graph <http://montefiore.org/edu#3ee75fbb416ce755cda2ce5db8500f71> { ?x ?y ?z .}}
```

Below the query, the results are displayed in a table with 1000 of 4995 results. The table has three columns: ?x, ?y, and ?z. The results show a list of medication administrations for a specific patient.

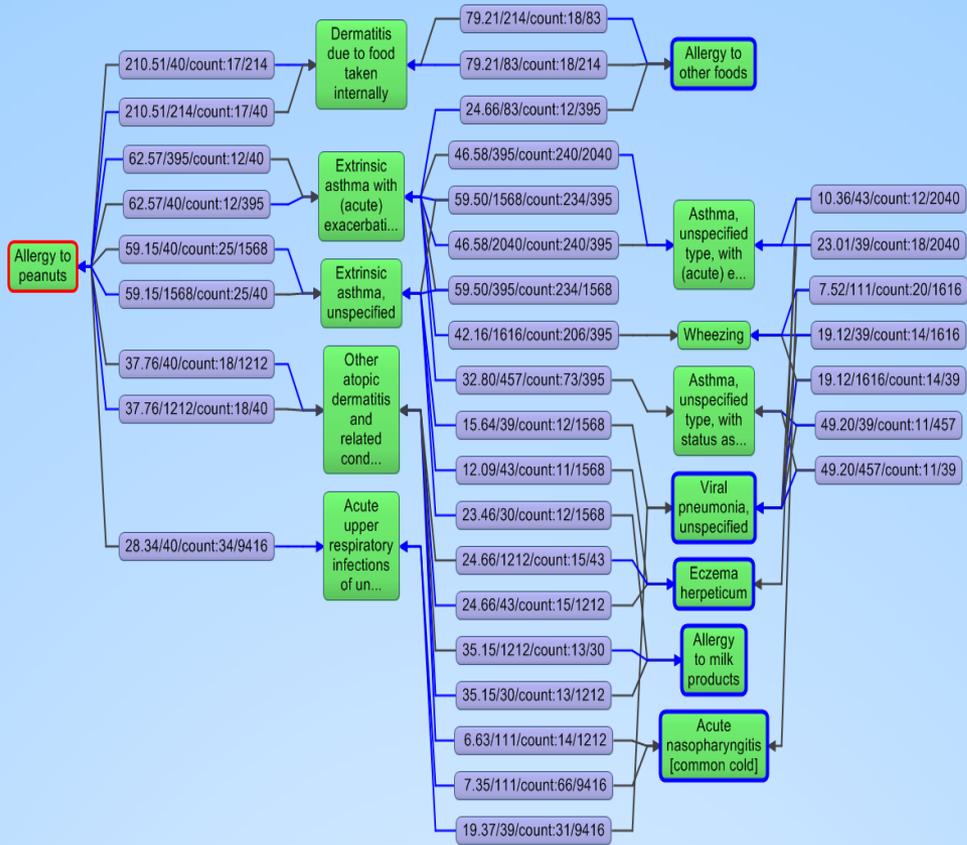
?x	?y	?z
3ee75fbb416ce755cda2ce5db8500f71	Source URI	19ee432e7e5ec8abde249a31f4dbe90e
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	8b67be581404ae12abcc711a8d7034f6
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	Abddb6650e58a25b96221e8349226fbc
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	1a4ac678816ad5a97670e23965a89237
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	948c2e084d2e5f8748b2e17889040777
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	A028d27ce67b670fa8f54dac775cd2e0
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	9659da4364ea10004bd7260133a8e162
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	F2834207b0dae2d41567567ed6803c1a
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	08c61bd64abd4b0f3938e036c2de75bb
3ee75fbb416ce755cda2ce5db8500f71	Medication Administration	5518e073dde2ed4a97d65108c7fecf3b

Simple example of putting output of analytics back in the Knowledge Graph

- Find interesting co-occurrences (oddsratios) and make them 'graph' discoverable
- For every disease we can show you the top 5 most related ones
- Based on your EMR we can easily find all your risk areas. Some are very surprising.

Patient Population		1,802,464		
		Ingestion Dermatitis		
		TO+	TO-	
Peanut Allergy	FROM+	544 (5)	736	1,280
	FROM-	6304	1,795,424	
		6,848		
Odds Ratio	210.51			
95% CI Lower	187.91			
95% CI Upper	235.82			

From5
To5
Co Occurrence
Concept



Take away

- Building knowledge graph is great fun
- Semantic technology ensures that your Knowledge Graph can play with other knowledge graphs.

The END
questions?