



Determining the Most Influencing Medical Conditions in MEDPRAT's SIN Directed Graph

Crew Health and Performance
Probabilistic Risk Assessment Project
(CHP-PRA)

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Background and Introduction

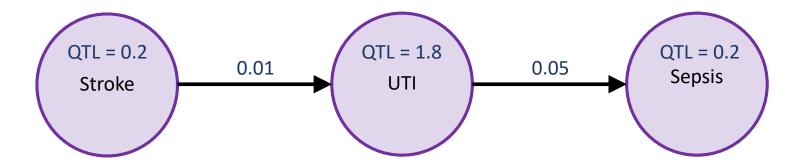


SIN Graph



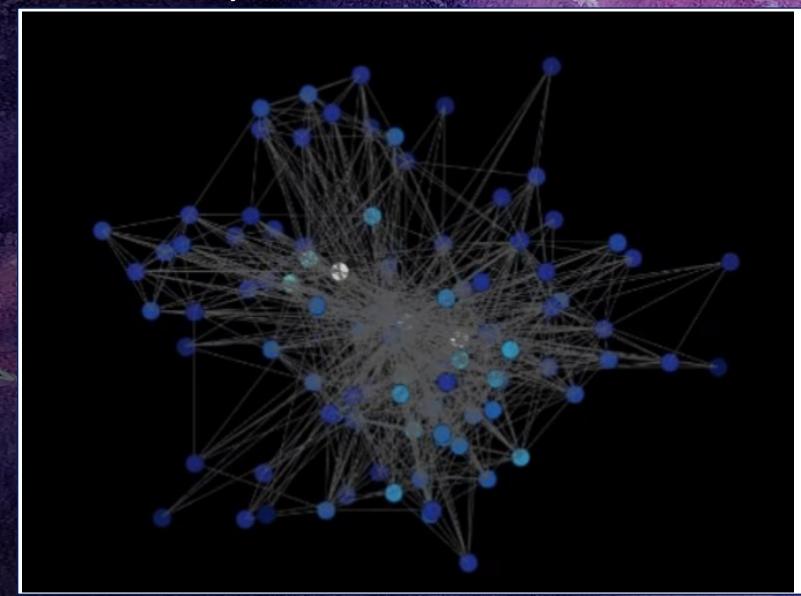
The Susceptibility Inference Network (SIN)

- Part of MEDPRAT, an event driven, time dependent, stochastic tool
- Subject matter expert informed prototype
- Mathematical data structure to quantify relationships between medical conditions:
 - o nodes: medical conditions
 - o edges: directed, progression from one condition to the next
 - o node weights: a statistic to measure the stand-alone risk, typically Quality Time Loss (QTL) or Loss Of Crew Life (LOCL)
 - o edge weights: probability of progression to the next condition



SIN Graph





- 99 medical conditions
- 1078 direct connections
- Anxiety, Atrial Fibrillation/ Flutter, Medication Overdose/Adverse Reaction responsible for 273 of the connections



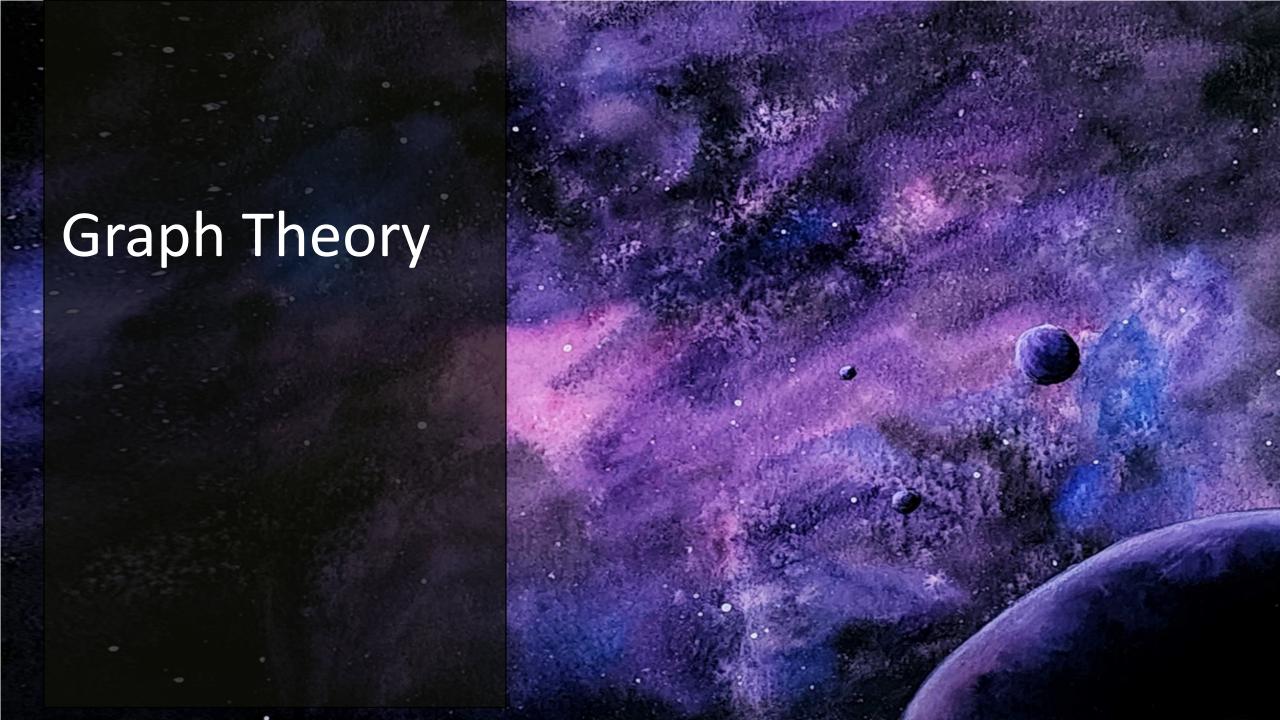
Centrality Measures



• Our goal 1: compare how much each medical condition can progress through the network

- Measure of importance of a node
 - Local measures: degree centrality
 - Global measures: betweenness, eigenvector, Katz, PageRank Centrality

We use: Katz Centrality



SIN Adjacency Matrix

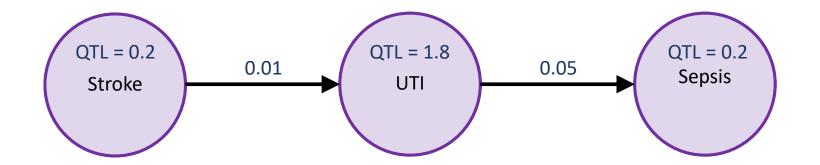


SIN adjacency matrix

Stand-alone risks

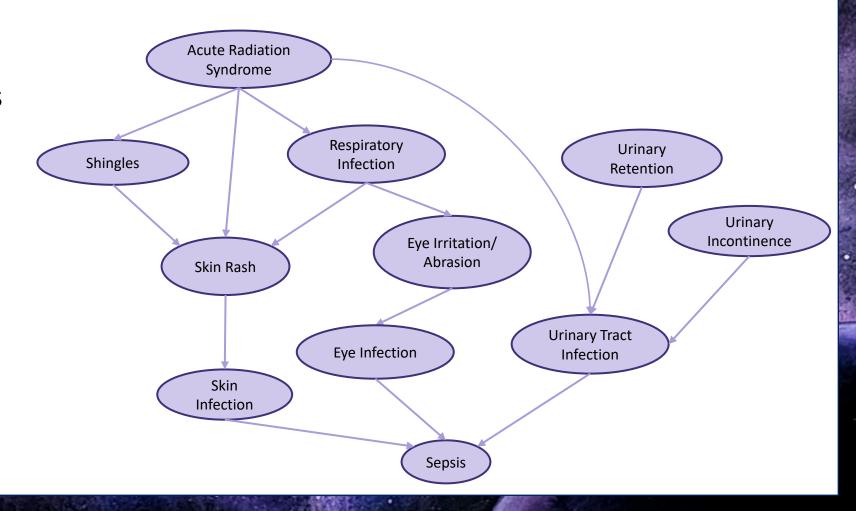
Stroke
$$\begin{bmatrix} 0.2 \\ \end{bmatrix}$$

$$W = \begin{bmatrix} 0.2 \\ \end{bmatrix}$$
Sepsis $\begin{bmatrix} 0.2 \\ \end{bmatrix}$



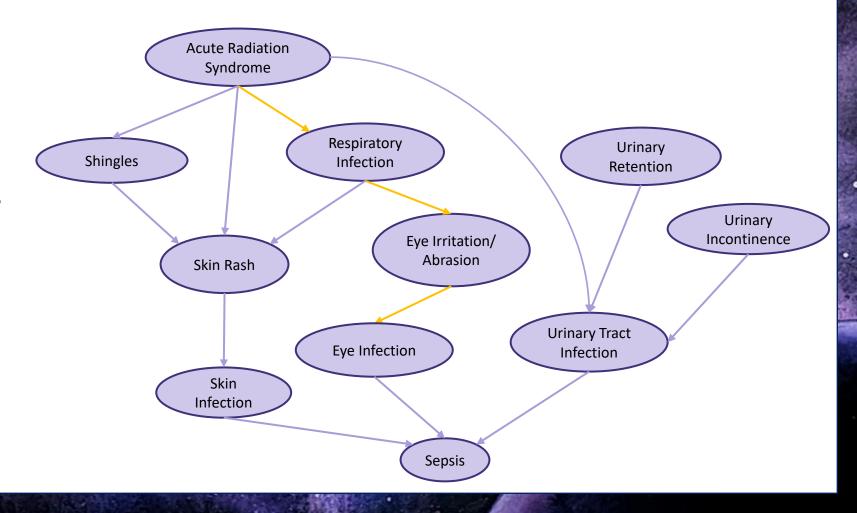


- ullet Walk of length k through k edges
- Weight of walk= product of edge weights



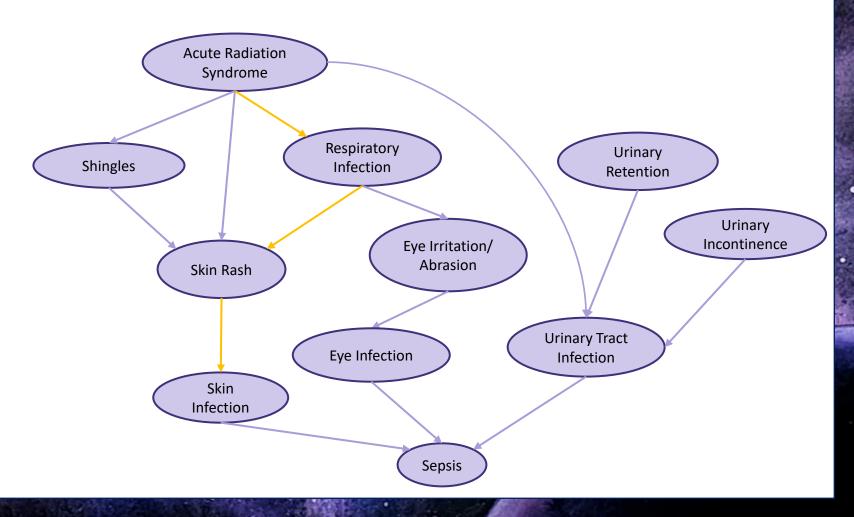


- Walk of length 3
- Weight of walk= product of edge weights



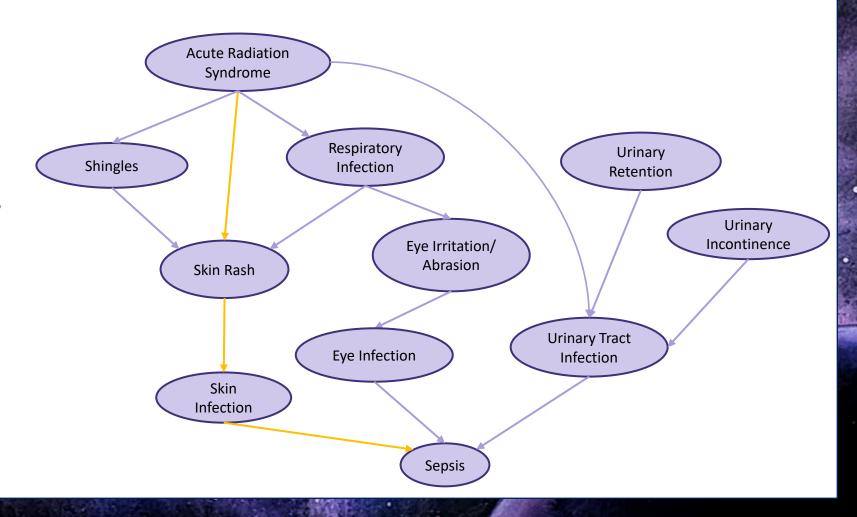


- Walk of length 3
- Weight of walk= product of edge weights



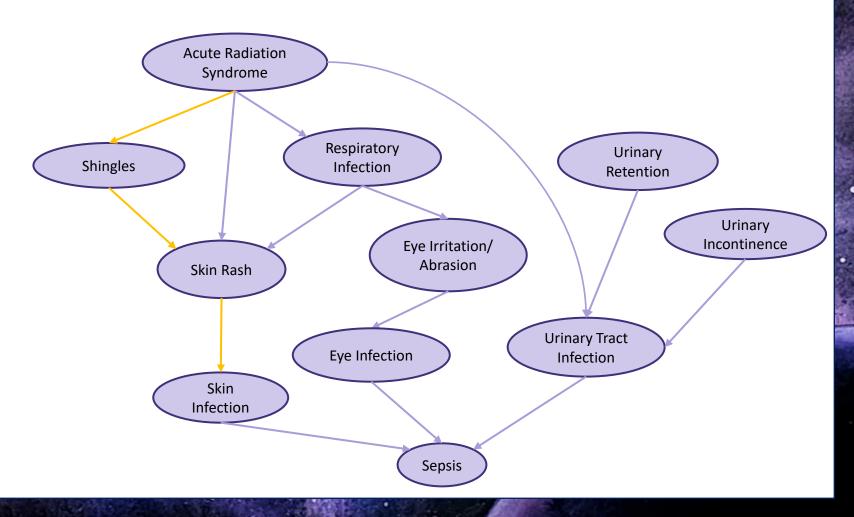


- Walk of length 3
- Weight of walkproduct of edge weights



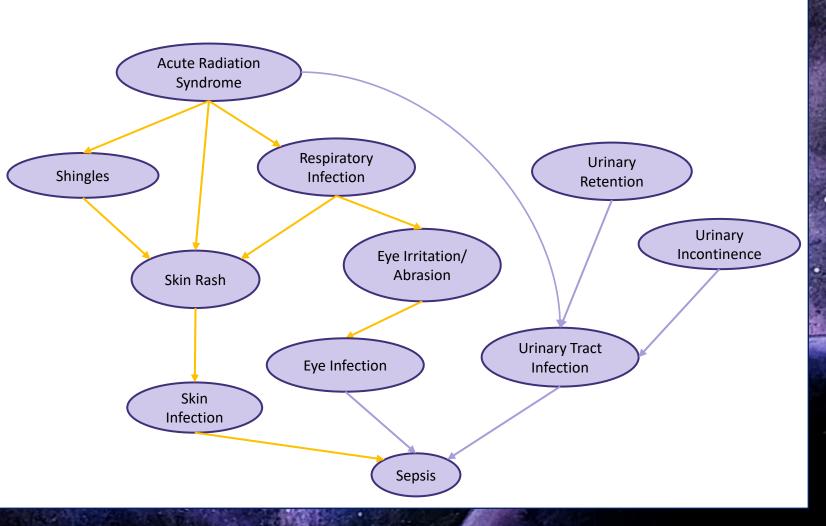


- Walk of length 3
- Weight of walk= product of edge weights





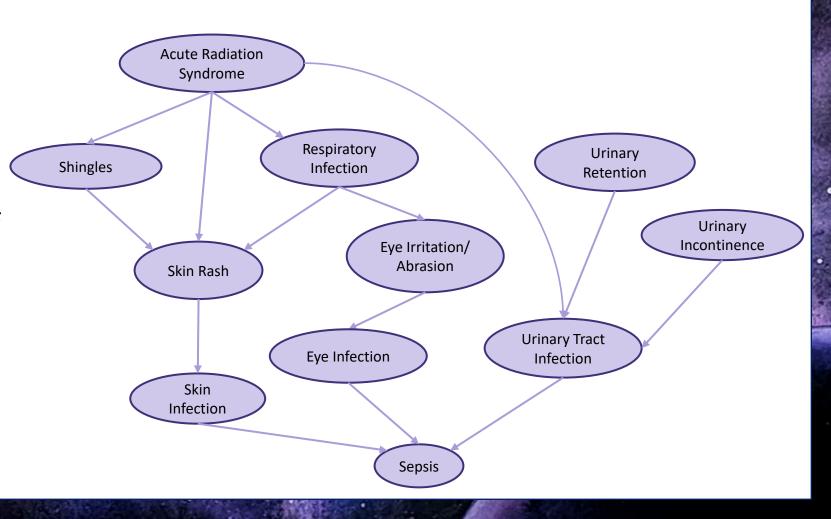
- All walks of length 3
- Weighted number of walks of length 3
 - = sum of weighted number of all walk of length 3





For adjacency matrix A

- Weighted number of walks of length k= $A^k \cdot 1$
- Sum of weighted number of walks of any length $=(A + A^2 + A^3 + \cdots) \cdot 1$



Katz Centrality



ullet The Katz Centrality of vertex $oldsymbol{i}$ in a graph with adjacency matrix A:

$$C(\alpha)_{i} = (\alpha A + \alpha^{2} A^{2} + \alpha^{3} A^{3} + \cdots) \cdot W$$
$$= \left((I - \alpha A)^{-1} - I \right) \cdot W$$

• Katz parameter α to penalize longer walks

• Criteria to choose α not well defined in the literature

Katz Parameter



• Our goal 2: give guidance to the choice of α

 Based on the maximum path length of interest L when navigating the network

• Takeaway: for maximum path length L and error tolerance ϵ , choose α that satisfies

$$\left[\log_{\alpha\|A\|_{2}}\left(\frac{\epsilon}{\|C(\alpha)\|_{2}}\right)\right] = L$$

Our propositions and theorems:

Proposition 1 Let $C \mid and C'$ be two centrality measures on a network N. If $||C - C'||_{\infty} < \epsilon$,

then C and C' ϵ -agree.

Lemma 2 (Absolute Error Tolerance) Let $p \in \{1, 2, \infty\}$ and $\alpha \in (0, 1/\rho)$. Then

$$||C(\alpha) - C(\alpha, \ell)||_{\infty} \le (\alpha ||A||_p)^{\ell} ||C(\alpha)||_p := \epsilon_{\ell}.$$

Lemma 3 (Error Tolerance Guarantee) Let $p \in \{1, 2, \infty\}$, $\alpha \in (0, 1/\|A\|_p)$ and $\epsilon > 0$. If

$$\ell > \log_{\alpha \|A\|_p} \left(\frac{\epsilon}{\|C(\alpha)\|_p} \right) := L$$

then $||C(\alpha) - C(\alpha, \ell)||_{\infty} < \epsilon$.

Corollary 4 (Relative Error) Let G be a graph, A the adjacency matrix of G, $\alpha = \alpha_0/\|A\|_2$, and $\epsilon = \epsilon_0\|c_\alpha\|_2$ where $\alpha_0, \epsilon_0 \in (0,1)$. Then α -Katz centrality and (α, ℓ) -Katz centrality are ϵ -tolerant whenever $\ell > \log_{\alpha_0}(\epsilon_0) = L$.

Corollary 5 (Local Katz Centrality) Let G be a directed graph, $\alpha \in (0, 1/\|A\|_2)$, $\epsilon > 0$, and L be as defined in Lemma 2. Let H be the subgraphs formed by the L-hop neighborhood of v in G. Then the α -Katz centrality of $v \in G$ is within ϵ of the α -Katz centrality of $v \in H$.

Application to SIN



Example Mission



Given short mission details:

- 1. 6 months long
- 2. L = 6
- 3. Error tolerance $\epsilon_0 = 0.001$

Use calculated $\alpha = 0.37$

Rank (QTL)	Condition	Rank (LOCL)	Condition
1	Acute Radiation Syndrome	1	Acute Radiation Syndrome
2	Anaphylaxis	2	Burns Secondary to Fire
3	Allergic Reaction	3	Appendicitis
4	Toxic Exposure	4	Urinary Tract Infection
5	Abdominal Injury	5	Sudden Cardiac Arrest

Example Mission



Given long mission details:

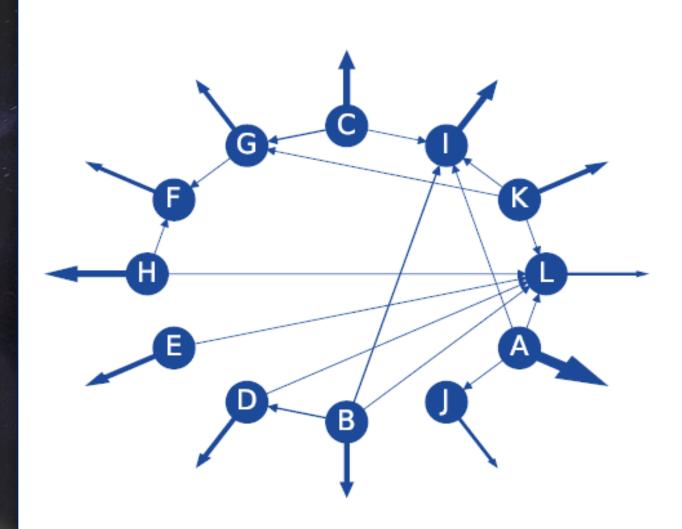
- 1. 3 years long
- 2. L = 15
- 3. Error tolerance $\epsilon_0 = 0.001$

Use calculated $\alpha = 0.80$

Rank (QTL)	Condition	Rank (LOCL)	Condition
1	Acute Radiation Syndrome	1	Acute Radiation Syndrome
2	Anaphylaxis	2	Burns Secondary to Fire
3	Toxic Exposure	3	Appendicitis
4	Allergic Reaction	4	Urinary Tract Infection
5	Abdominal Injury	5	Sudden Cardiac Arrest

Subnetwork of the SIN with Highest Centrality Nodes





For α =**0**.37

- 12 top ranked conditions in the SIN in contributing to the QTL risk
- Thickness of edges pointing outwards reflecting total edge weight towards nodes in the remainder of the network
- Node A in first rank: Acute Radiation Syndrome

Conclusions and Limitations



Conclusions and Limitations



Conclusions

- The nodes Katz centrality scores depend on Katz parameter α .
- We recommend an α based on the depth to navigate in the network.
- Medical conditions are ranked differently for different mission profiles, based on the number of progressions of concern during each mission.

Limitations

- α is bounded by the inverse of the spectral radius of the graph's adjacency matrix which limits the range of α to consider.
- We can get smaller margin of error computationally than analytically.

