

Milestone TWO “:” Exploratory Network Analysis of Clinical Interactions in the ED

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^a*Find the GitHub repository for this Milestone at
https://github.com/tommyflynn/Flynn_N741_Project/tree/master/Flynn_Project*

Abstract

Patient acuity in the Emergency Department is triaged at the beginning of the care process using the Emergency Severity Index (ESI). Although the ESI is widely used and accepted as a validated predictor of ED resource consumption, its predictive power has limitations that can negatively impact patient flow and safety. An objective measurement of individual resource consumption, that passively observes and calculates relative patient needs, would allow nurses and administrators to make better decisions with staffing and patient assignments. This study proposes an approach to continuous passive patient acuity monitoring using radio-frequency identification (RFID) technology and network analysis algorithms to define and potentially predict changes in patient acuity. The first step in this process is to determine how acuity, as it is currently measured, correlates to the network position of patients in the clinical interaction network. A strong correlation between network position and acuity is expected, and a change in network position is presumed to reflect a change in patient condition.

Research Question & Specific Aims

Can network analysis of clinical interactions between patients and staff provide insight into the complex Emergency Department patient care process?

Aim 1: Create a network graph from the data. (Baby steps)

Aim 2: Create a visually engaging story of ED network development

Background & Objectives

In an Emergency Department (ED), care is delivered over a network of face to face human interactions. Patients interact with registration staff, then a triage nurse who may decide to discuss the patient with a provider, the provider may then interact directly with the patient, and so on. In this way, the network grows over time, creating a web of care that may correlate with the amount and quality of care delivered to individual patients. - **The purpose of this study is to explore the network of clinical interactions that take place in the Emergency Department and describe the relationship between those network variables and patient acuity.** To study this relationship, received permission to analyse existing data that includes the following; the frequency and duration of all face-to-face interactions (patients, providers, nurses, technicians, & administrators) that occurred in the ED for 81 12hr shifts, the location of those interactions, and individual patients' medical and demographic characteristics including acuity, chief complaint, gender, age, arrival mode, and disposition. The network structural characteristics will be assessed in relation to the industry standard acuity measure, the Emergency Severity Index (ESI), and potential confounding variables. Using this data will require specific knowledge of the R statistical packages, network analysis, and data science. See Tables 1-4 for my learning goals with respective action items, timeline, and outcomes.

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Data

This study applies a secondary data analysis design due to the exploratory nature of the research aims. Data was made available with permissions from the originating research team. The purpose of the original study was to describe contact characteristics between patients and staff in the ED of a busy urban hospital to inform cross-infection control measures. Data were collected using a radio-frequency identification system that triangulated patient and staff (nurses, providers, and ancillary staff) locations within the ED at Emory University Hospital Midtown. The end result of the RFID collection system is a data set with every participant's location for the duration they were in the ED. With the location of every individual in the ED, we can determine when, and where, two individuals came into within 1 meter of one another while being in the same room. And that is how we define an interaction. Data for this secondary analysis were collected using a prospective, longitudinal, observational design with a random sampling of one day shift and one night shift per week for one year, July 1, 2009 to June 30, 2010. This strategy was chosen to minimize sampling bias related to seasonal or weekly fluctuations in census, acuity, and ED staffing changes. Although a total of 104 shifts were observed, the original research team retained only 81 shifts for reasons related to issues with the RFID system and study staff sick leave. (Lowery-North et al. 2013)

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## [137]	"Respiratory"	"Unexplained_death"
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## [19] "Chief_Complaint" "Acuity" "Arr_Mode"
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```

Data Wrangling:

Now that I have the data, it turns out that the files are too large (1.5G) for my **little** computer. Therefore, I will move forward with a previous subset of the data that was given to me earlier in the year. Although this is not the full data set, it will suffice while I learn the process needed to clean, explore, analyze, & report on the data with R and Rmarkdown.

Analysis Plan

Exploratory Analysis

Descriptive statistics of the network data as well as patient demographic data will be evaluated for assumptions of normality. The data will be skewed in certain predictable ways due to the observed patient populations. The distribution of study subject demographics will be described in tabular format, noting irregularities and potential sources of error.

Variables used for final analysis:

Network Variables > - *Patient eigenvector centrality* (dependant variable of interest) > - Network density > - Network clustering coefficient > - Network diversity scale??

Staff variables > - Title (RN, MD, staff)

Patient variables > - *Acuity* (ESI, independent variable of interest) > - Comorbidities (index) > - Gender > - Age > - Race > - Ethnicity > - Arrival mode (ambulance v. walk-in) > - Education (if available) > - Disposition (admission v. discharge) > - Length of stay (common measure of quality in the literature used for comparison) > - Time before first provider contact (common measure of quality in the literature used for comparison)

Analysis

We start with descriptive statistics of all relevant variables for assumptions of normality. Multiple linear regression will be used for the final analysis to assess the correlation between patient acuity and patient centrality. Relationships will be evaluated visually (see below) as well as statistically to an alpha of 0.05.

Schedule

-Milestone 1: February 14th, 2018 -Milestone 2 UPDATE: March 28th -Milestone 2: April 4th, 2018 -Final Proposal: April 26th, 2018

Results

Results will be discussed with the visual supplementation of network graphs. This allows the reader to understand concepts that may be difficult to grasp through text alone.

Discussion

Allocating staff resources in an Emergency Department is an ongoing challenge. How can these results begin to offer solutions to ED staff and patient management?

What were my primary limitation (both expected and unexpected)?

Conclusion

Did I meet my learning objectives? How would I design a better study next time?

References

Lowery-North, Douglas W., Vicki Stover Hertzberg, Lisa Elon, George Cotsonis, Sarah A. Hilton, II Vaughns Christopher F., Eric Hill, Alok Shrestha, Alexandria Jo, and Nathan Adams. 2013. "Measuring Social Contacts in the Emergency Department." Journal Article. *PLoS ONE* 8 (8): e70854. doi:[10.1371/journal.pone.0070854](https://doi.org/10.1371/journal.pone.0070854).