

The impact of Gender, Medical History and Vital Status on Emergency Visits and Hospital Admissions: A Remote Patient Monitoring Case Study

Catherine Inibhunu¹ *Member, IEEE*, Adrian Schauer²,
Olwen Redwood³, Patrick Clifford⁴ and Carolyn
McGregor^{1,5}, *Senior Member, IEEE*

¹University of Ontario Institute of Technology, Oshawa, Ontario, Canada

²Alaya Care, Toronto, Canada

³We Care, Toronto, Canada

⁴Southlake Regional Health Centre, Newmarket, Ontario, Canada

⁵University of Technology Sydney, Ultimo NSW, Australia



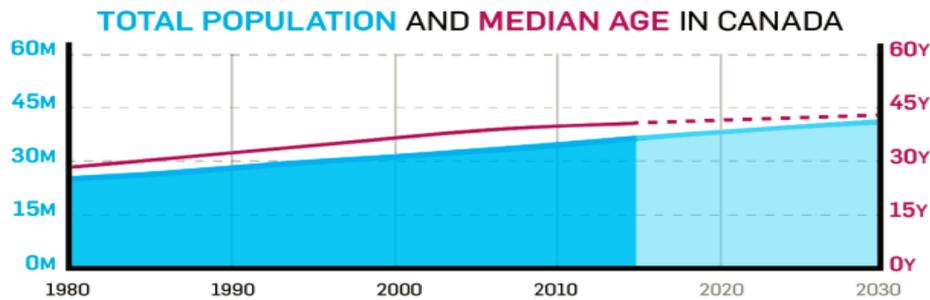
**Presented at IEEE LSC 2017:
Sydney, Australia, Dec 13 – 15, 2017**

Agenda

- Research Problem
- Research Objective
- Overview of RPM Program
- Methods
- Results
- Key Findings
- Conclusion

Research Problem

➤ Aging Population & Associated HealthCare Costs

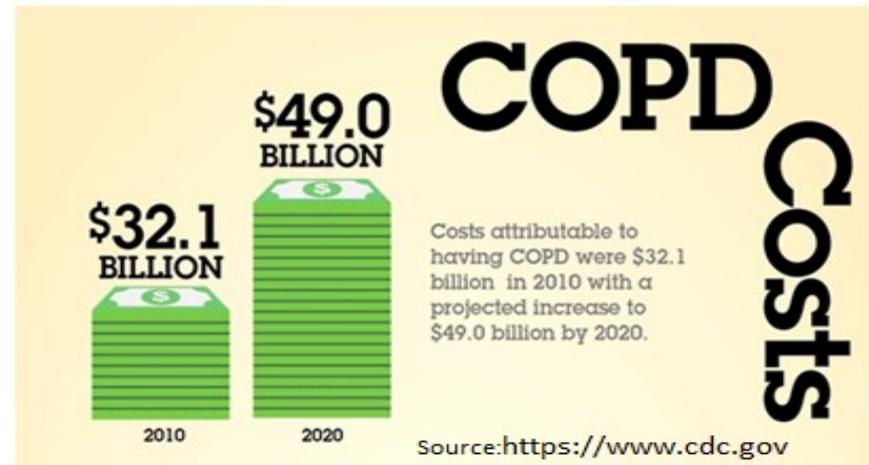


CANADIANS AGED 65 AND OLDER:

2005: 1 IN 8  2030: 1 IN 4 

Source: <http://www.ctvnews.ca/health>

- ❖ In 2012/13, Canadians over 65 accounted for 78% of the most expensive type of hospital stays: COPD, pneumonia and HF (CIHI).
- ❖ Patients with COPD: Highest rates of Hospital Readmissions, return within 7 days to ER Visit.



Key Questions?

- ❖ What are the contributing factors to lengthy hospitalizations and multiple Emergency Visits on patients with COPD and HF?
- ❖ Can identification of such facts lead to reduction on healthcare costs as well as improved outcomes for the patients.

Research Objective

- ❖ Utilize Remote Patient Monitoring (RPM) Program
- ❖ Demonstrate that Predictive analytics applied on patient data captured remotely can help identify risk factors to lengthy hospitalization and multiple ER Visits.
- ❖ Key Metrics Evaluated: Impact of gender and medical history on ER visits and hospital admissions

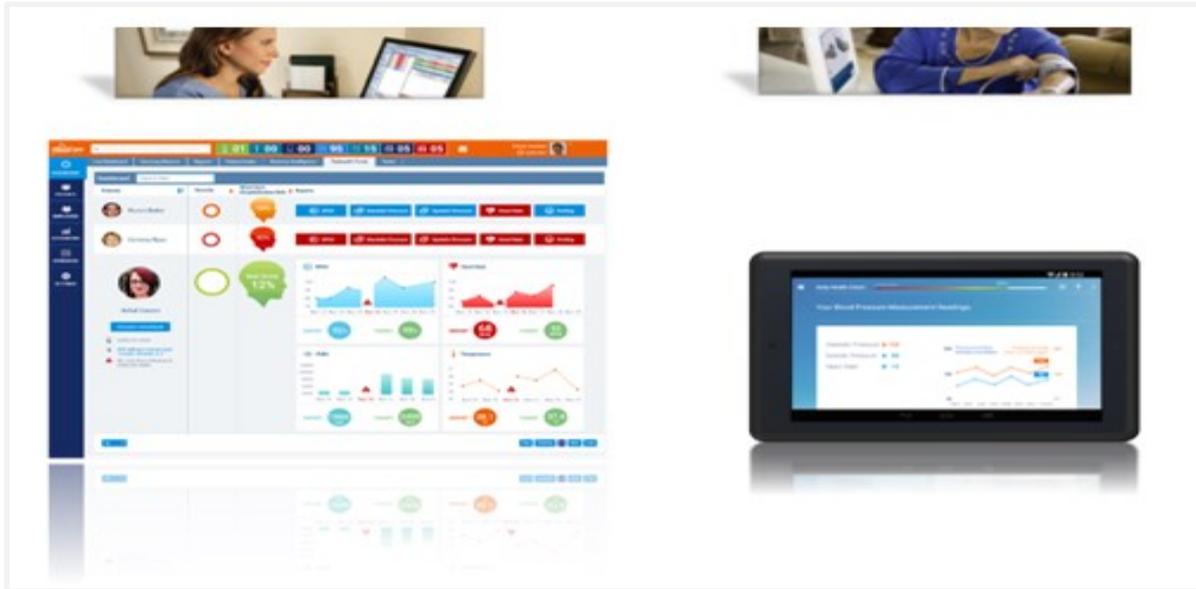
❖ Partners



Overview of RPM Program

Program Goal:

- ❖ Reducing Hospital Admissions and Emergency Department Visits for Chronically ill patients using Remote Patient Monitoring and Telehealth Tools



Facilitated by:

- ❖ Patient Monitoring
- ❖ Data Collection
- ❖ Analysis
- ❖ Action

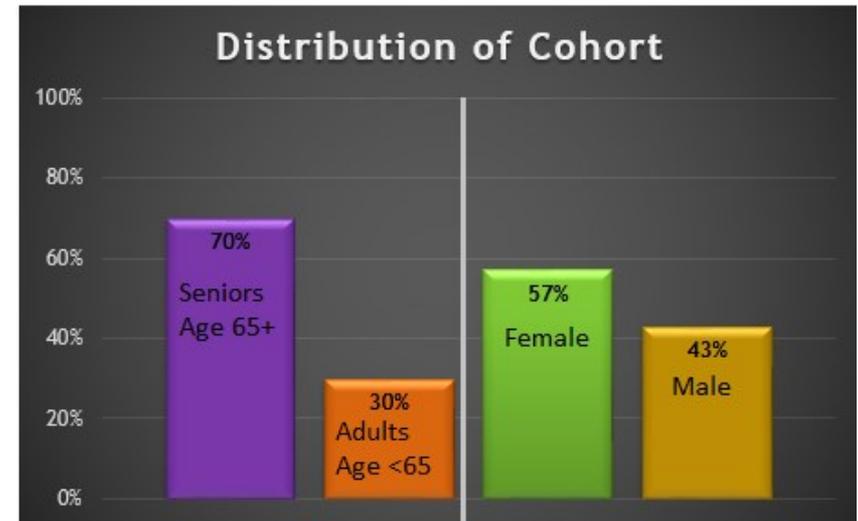
Methods

➤ Data Details

- ❖ A subset of de-identified dataset collected from patients participating in the RPM programs in 2016-17.
- ❖ Data elements included chronic disease, age, sex, hospitalization details, emergency room visit details and clients vital status.
- ❖ Data Preparation: To facilitate the analysis: Cleaning, Linking and Standardization

➤ Summary Statistics

- ❖ 69, STD: 17.6, Min 20, Max 97, (N=84)
- ❖ 14% more Female Clients than Males

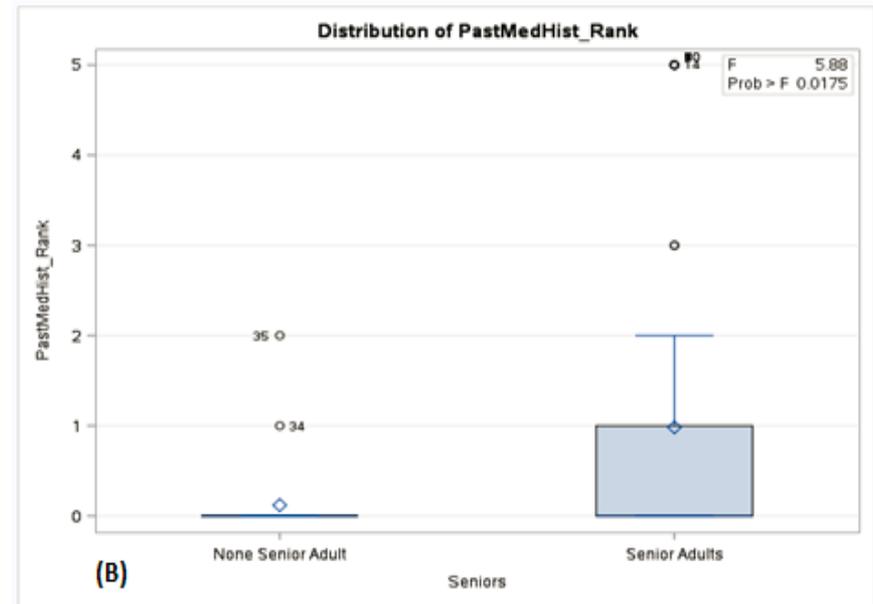
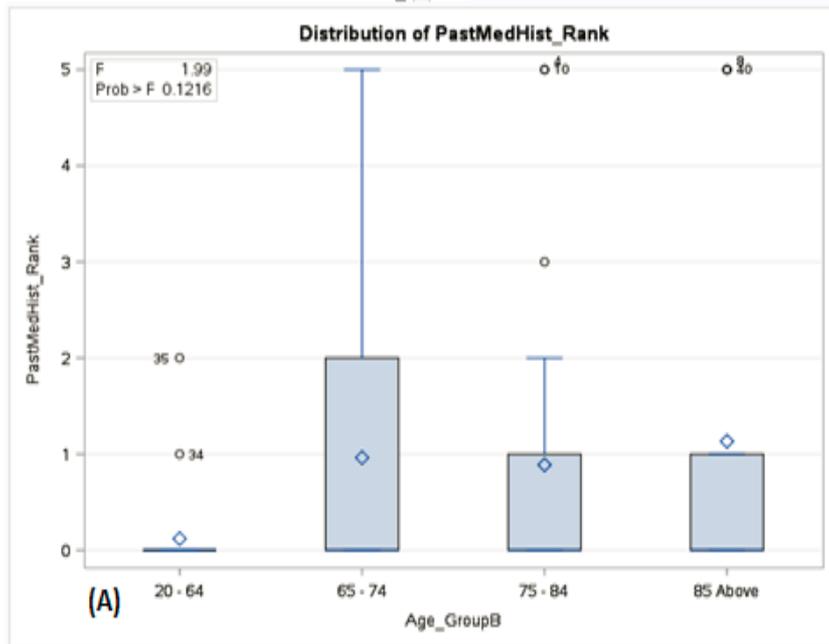


➤ Predictive Analytics

- ❖ Probabilistic Analysis
- ❖ Correlation Analysis on patient attributes to hospital admissions

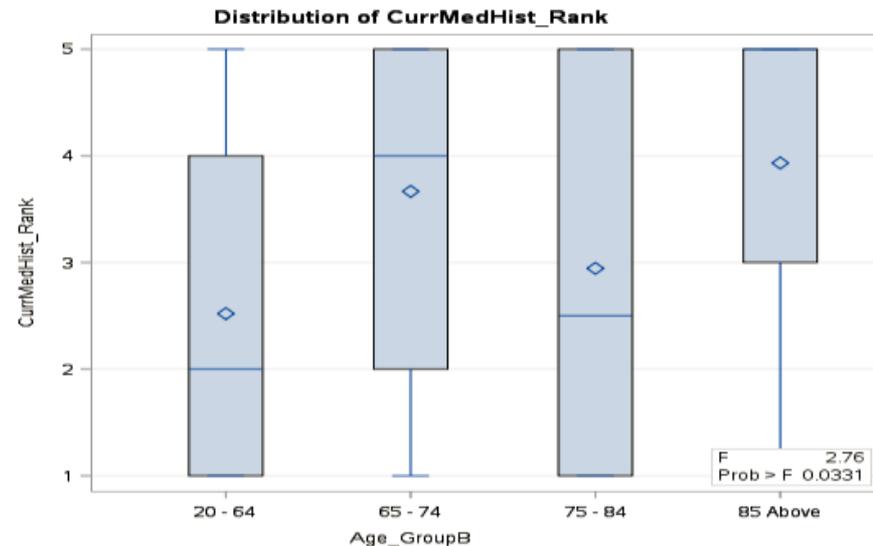
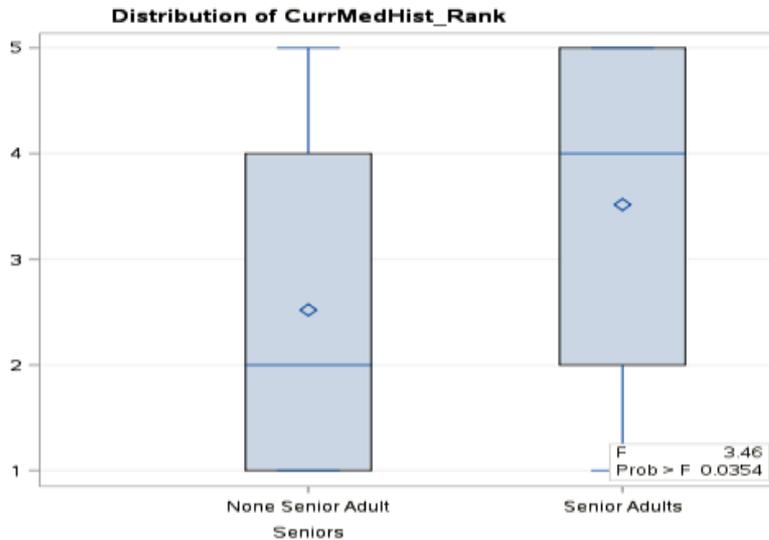
Results

❖ Variation on Client Medical History



- ❖ The probability of having past medical records based on age is statistically significant
- ❖ The older the patient the higher the presence of more than one comorbidities ($p=0.0175$)

Distribution of Current Medical by Age



- ❖ Statistically significant number of seniors adults with current medical history ($p=0.0354$).
- ❖ A drill down on senior adults indicates a statistically significant number on clients aged 85 and over, ($p=0.00331$).
- ❖ No indication that past and current medical history varies by gender

Hospital Admissions

- ❖ 20 hospital admissions, length of stay 2 to 11 days.
- ❖ 10 unique patients, 60%, 2 or more hospital admissions
- ❖ Exasperation of COPD was the most common reason for of the hospitalization (64%).

Correlation of Hospital Admissions by Gender

Gender = Male

Pearson Correlation Coefficients, N = 45 Prob > r under H0: Rho=0				
	PastMedHist_Rank	CurrMedHist_Rank	Allergy_Rank	hadmitflagA
PastMedHist_Rank	1.00000	0.28391 0.0588	-0.18180 0.2320	0.54218 0.0001
CurrMedHist_Rank	0.28391 0.0588	1.00000	0.27665 0.0658	0.25335 0.0931
Allergy_Rank	-0.18180 0.2320	0.27665 0.0658	1.00000	0.40786 0.0054
hadmitflagA	0.54218 0.0001	0.25335 0.0931	0.40786 0.0054	1.00000

- ❖ Statistical significant correlation for male clients on hospital admissions and past medical history ($p=0.0001$), allergies ($p=0.0054$).

- ❖ For female participants, no such correlation is found

Gender = Female

Pearson Correlation Coefficients, N = 53 Prob > r under H0: Rho=0				
	PastMedHist_Rank	CurrMedHist_Rank	Allergy_Rank	hadmitflagA
PastMedHist_Rank	1.00000	0.42006 0.0017	0.59503 <.0001	0.01317 0.9254
CurrMedHist_Rank	0.42006 0.0017	1.00000	0.46671 0.0004	-0.20971 0.1318
Allergy_Rank	0.59503 <.0001	0.46671 0.0004	1.00000	-0.12991 0.3539
hadmitflagA	0.01317 0.9254	-0.20971 0.1318	-0.12991 0.3539	1.00000

- ❖ However, there was statistically significant correlation between allergies and past medical history at ($p<0.0001$) on females

Correlation Analysis by Age, Gender

Age 75 to 84, no hospitalizations

- ❖ On males, strong correlation on past and current medical history that was statistically significant.
- ❖ On Females presence of allergies was associated with Past medical history

Age 85+ with Hospitalizations

- ❖ Strong correlation between past and current medical history on males, similar findings not found on females

Gender = Female

Age_GroupB=75 - 84

Pearson Correlation Coefficients, N = 10 Prob > r under H0: Rho=0			
	PastMedHist_Rank	CurrMedHist_Rank	Allergy_Rank
PastMedHist_Rank	1.00000	0.43072 0.2140	0.84095 0.0023
CurrMedHist_Rank	0.43072 0.2140	1.00000	0.58381 0.0784
Allergy_Rank	0.84095 0.0023	0.58381 0.0784	1.00000

Gender = Male

Age_GroupB=75 - 84

Pearson Correlation Coefficients, N = 5 Prob > r under H0: Rho=0			
	PastMedHist_Rank	CurrMedHist_Rank	Allergy_Rank
PastMedHist_Rank	1.00000	0.95258 0.0123	-
CurrMedHist_Rank	0.95258 0.0123	1.00000	-
Allergy_Rank	-	-	-

- ❖ Strong indication of differing features by Age and Gender
- ❖ Need careful evaluation/factorization of variables used in predictive modelling

Correlation on Client vital Status

- ❖ Statistical significant correlation between min, max blood pressure and weight ($P < 0.001$).
- ❖ There is also significant correlation between SPO2 and pulse rate.

Pearson Correlation Coefficients					
Prob > r under H0: Rho=0					
Number of Observations					
	ave_weight	ave_pulse	ave_spo2	ave_bpmin	ave_bpmax
ave_weight	1.00000 4171	-0.14714 <.0001 4109	-0.03855 0.0150 3984	0.06260 <.0001 3871	0.11982 <.0001 3873
ave_pulse	-0.14714 <.0001 4109	1.00000 4913	-0.14872 <.0001 4764	0.08549 <.0001 4601	-0.00891 0.5457 4603
ave_spo2	-0.03855 0.0150 3984	-0.14872 <.0001 4764	1.00000 4765	0.02042 0.1731 4454	-0.00970 0.5176 4456
ave_bpmin	0.06260 <.0001 3871	0.08549 <.0001 4601	0.02042 0.1731 4454	1.00000 4604	0.49093 <.0001 4598
ave_bpmax	0.11982 <.0001 3873	-0.00891 0.5457 4603	-0.00970 0.5176 4456	0.49093 <.0001 4598	1.00000 4606

Next research questions:

- ❖ What does this correlation indicate on cohort of patients with/without hospitalizations?
- ❖ Is there any temporal relationships in vital status leading up to an adverse event?

Key Findings

Participation in RPM

- ❖ Larger volume of female clients who participated in the program at 57% compared to 43% male.

Variation on Commodities

- ❖ Analysis indicates variations by age and gender on the existence of multiple medical conditions.

Probability of Having Medical Conditions

- ❖ A statistically significant indication that Senior adults age 65+ have a past medication condition ($p=0.0175$).
- ❖ A statistically significant indication on presence on current medical conditions on seniors aged 85+ ($p=0.0331$).

Correlation Analysis

- ❖ On **female clients**, a strong correlation on presence of allergies, current and past medical history, however these factors were **not correlated to hospital admissions**.
- ❖ On **male clients**, past medical history ($p=0.0001$) and presence of allergies ($p=0.0054$) all **strongly correlation to hospitalization**.
- ❖ **Vital status, statistically significant correlation** on average: weight vs blood pressure, pulse vs weight, pulse and SPO₂ ($p<0.0001$)

Conclusion

- ❖ There is need to understand the cohort of patients participating in telehealth programs using Analytics
- ❖ Potential to drive the necessary care needed leading to improved patient experience, reduction of cost of care and better outcome.
- ❖ Analytics facilitated by statistical quantification of patient attributes thus provide evidence on variation across many data points collected on those patients.

This paper provides

- ❖ Several dimensions of analysis that shows variations among patients age and gender on; **presence of past and current medical history, hospitalization and distribution on clients vital status.**
- ❖ **In future works, we will perform further analysis to understand if hospitalization can be explained by the correlation seen in the client vital status prior to admission event** as opposed to analysis on the whole timeframe when clients participates in the study.

References

- [1] S. L. Gorst , C. J. Armitage, S. Brownsell and M. S. Hawley, "Home Telehealth Uptake and Continued Use Among Heart Failure and Chronic Obstructive Pulmonary Disease Patients: A Systemic Review," *Annals of Behaviour Medicine*, vol. 43, no. 3, pp. 323-336, doi: 10.1007/s12160-014-9607-x, 2014.
- [2] B. W. Ward, J. S. Schiller and R. A. Goodman, "Multiple Chronic Conditions among US Adults: A 2012 Update," *Preventing Chronic Disease*. DOI: <http://dx.doi.org/10.5888/pcd11.130389>, vol. 11, 2014.
- [3] We Care and Alaya Care, "Better Technology Better Outcomes: The Effects of Machine Learning Powered Remote Patient Monitoring on Home Care," 2013. [Online]. Available: <http://www.alayacare.com/wp-content/uploads/2015/01/Machine-Learning-White-Paper-1.pdf>. [Accessed August 2017].
- [4] CIHI, "Health Care in Canada, 2011. A focus on Seniors and Aging," 2011. [Online]. Available: https://secure.cihi.ca/free_products/HCIC_2011_seniors_report_en.pdf.