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2017

Internal Medicine

Keroles Hakem

Virginia Commonwealth University

Robert Trachy

Virginia Commonwealth University

Khanh Tran

Virginia Commonwealth University

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Internal Medicine

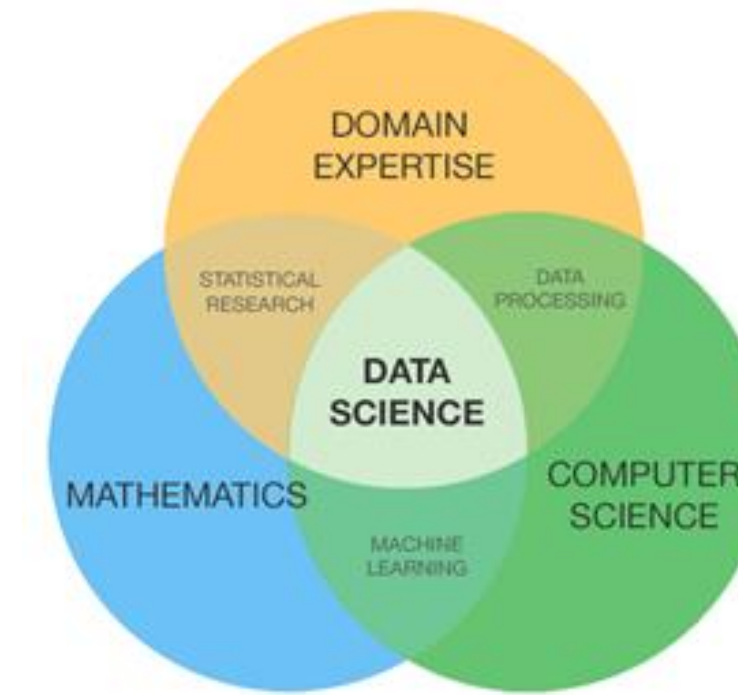
Project number: CS-318 | Team members: Kerolos Hakem, Robert Trachy, Khanh Tran | Faculty adviser: Bartosz Krawczyk, PhD; Vimal Mishra, MD

Objective

Using machine learning, predict the length of stay of patients for MCV hospital based on the different characteristics of a patient, e.g., method of arrival. Design a web app for the end users based on the final predictive models.

Approach

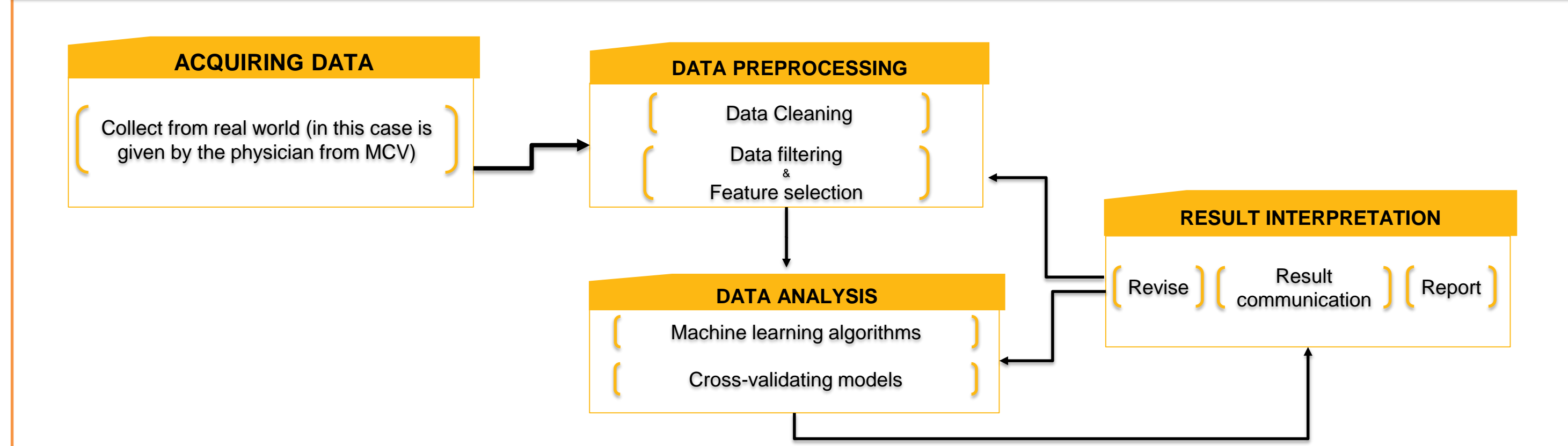
- Predictive models are developed utilizing:
 - Machine learning
 - Statistical methods
 - Professional input from domain experts.
- Machine learning techniques used include:
 - Supervised learning
 - Regression models (linear, ridge, lasso)
 - Cross-validation



Data Overview

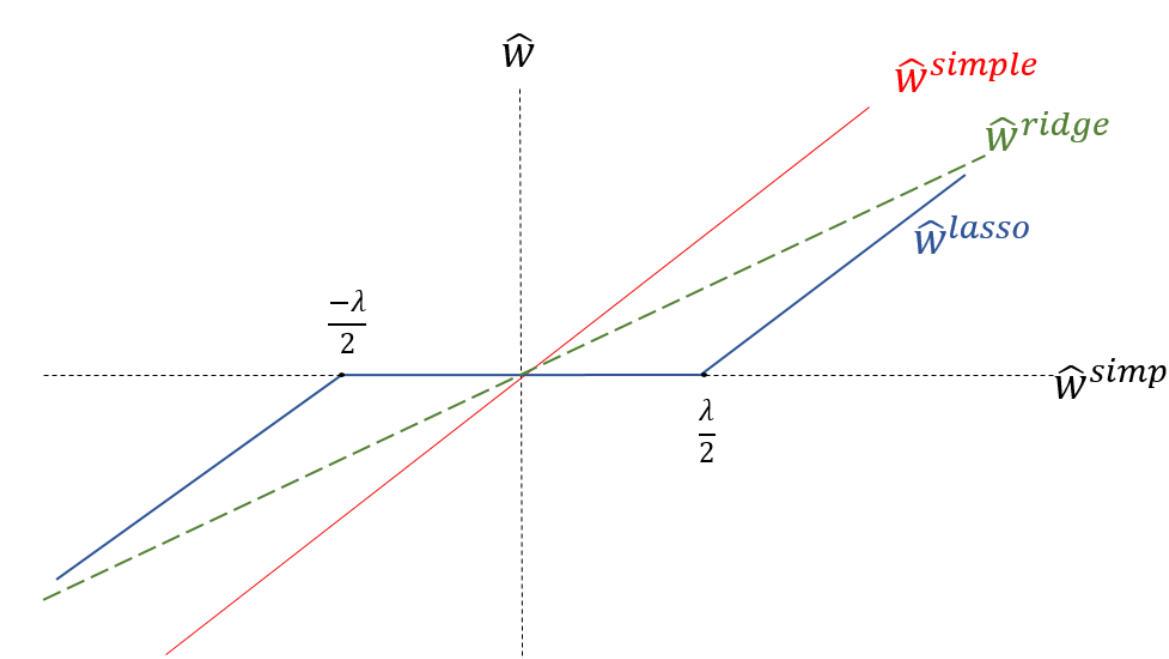
- Collected from MCV patients of the past 4 years
- Contained over 130,000 patients and 66 features
- Features include:
 - Clinical characteristics: primary diagnosis, universal disease group
 - Facility characteristics: bed category, admit unit, service provided, discharge disposition
 - Socioeconomic factors: Admit source, insurance, method of payment

Method



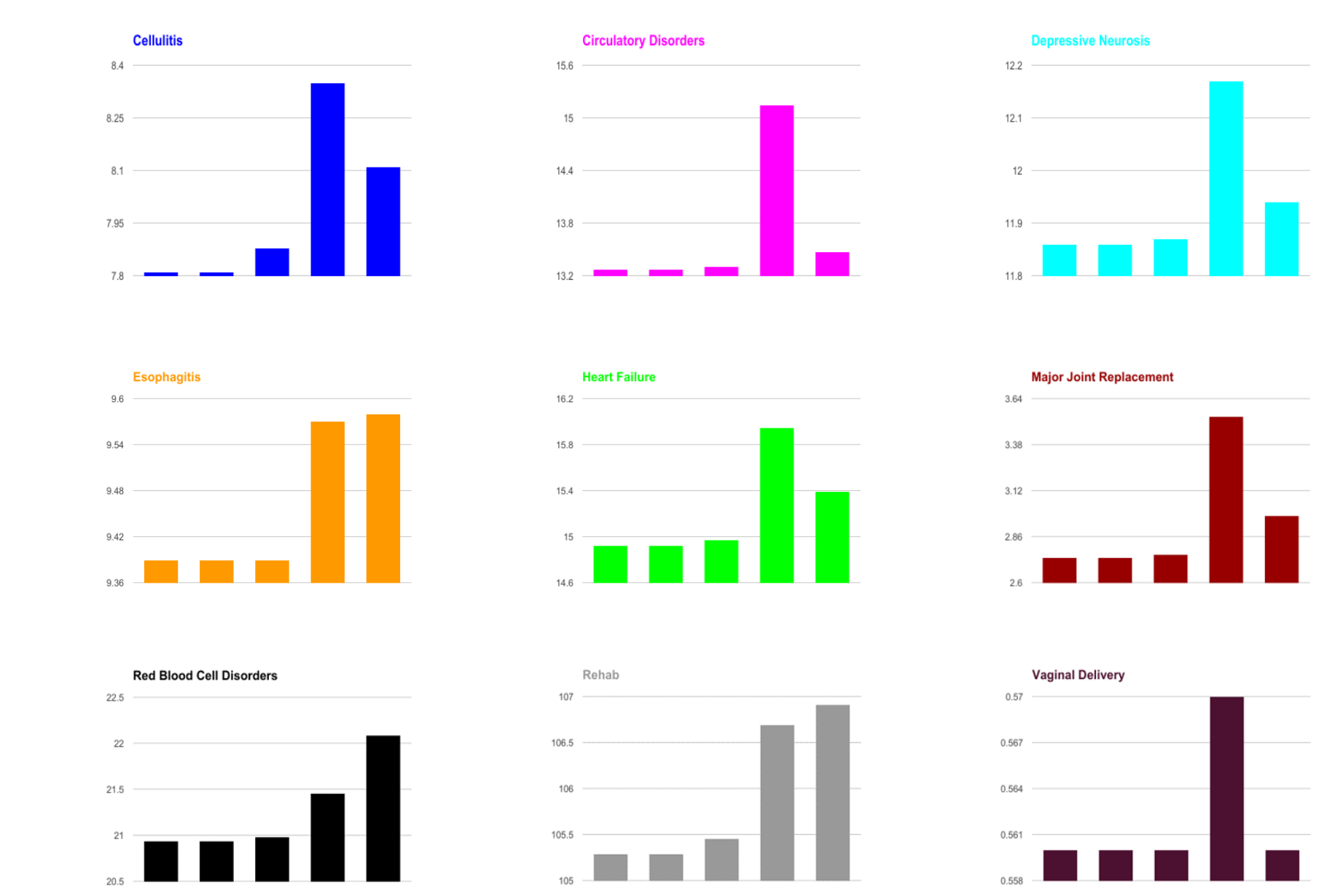
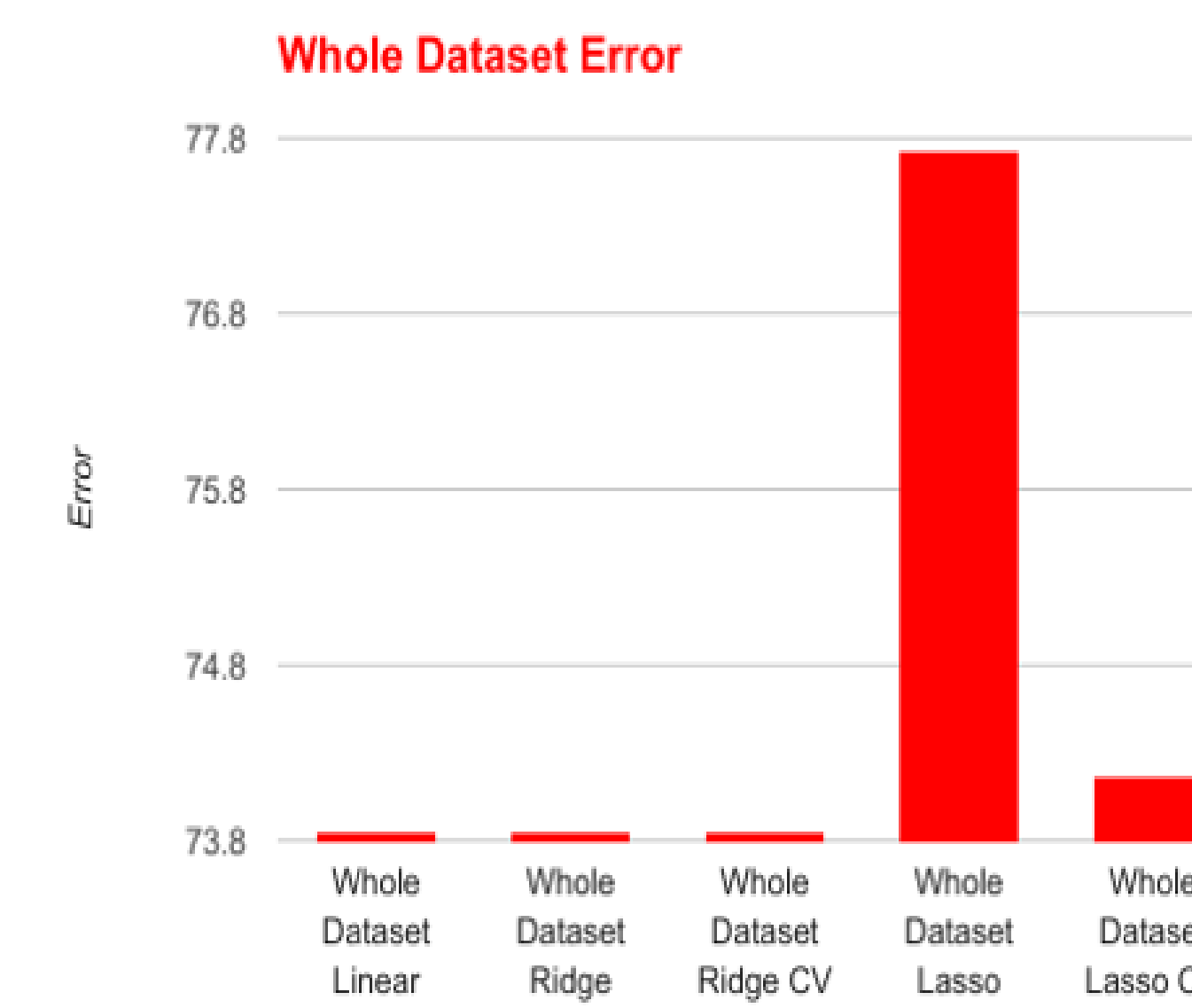
Linear vs. Ridge vs. Lasso Regression

- **Linear Regression:** finding set of weights that would minimize the overall error between the predicted and actual outcome using gradient descent
- **Ridge Regression:** using L2 regularization to shrinkage weights, i.e., adds penalty equivalent to **square of the magnitude** of coefficients
- **Lasso Regression:** using L1 regularization that would exclude unimportant features from model, i.e., adds penalty equivalent to **absolute value of the magnitude** of coefficients.



Result & Prototype

The final data consisted of 9 features: *admit source, primary insurance, discharge disposition, admit unit, iso result, icu order, stepdown order, general care order, and age*. Ten different models for the top ten most frequent diagnoses and one model for the whole data set have been built. Data trained with and without 10-cross-validation. Models are evaluated by the *mean square error* metric.



Prototype:

- A web application is developed with Angular 2 framework:
 - Friendly user interface that allows users to easily input information
 - Form validation to ensure valid input
 - Return an estimate of length of stays according to user input
 - Report statistical summary of each model including weights, errors, and variances

Future:

- In future research, more accurate prediction could be achieved with:
 - Specific dataset of a group or similar diseases
 - Larger dataset with more instances and potential factors
 - Diverse dataset from multiple hospitals' records

Conclusion:

Our research project showcased the potential of using machine learning regression methods for length of stay prediction.

pandas

$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$

