

Predicting readmission probability for diabetes inpatients

STAT 471/571/701, Fall 2017

Due: April 2, 2017 at 11:59PM

Instructions

- This project is due at **11:59pm on Sunday, April. 2, 2017.**
- It is an individual project, amounting to 15% of your final grade. See the *Collaboration* section at the bottom of this document.
- There is no single correct answer. You will be graded on the general quality of your work.
- The entire write up should not be more than 10 pages. You may put any supporting documents, code, graphics, or other exhibits into an Appendix, which is not counted in the 10 page limit.

Introduction

Background

Diabetes is a chronic medical condition affecting millions of Americans, but if managed well, with good diet, exercise and medication, patients can lead relatively normal lives. However, if improperly managed, diabetes can lead to patients being continuously admitted and readmitted to hospitals. Readmissions are especially serious - they represent a failure of the health system to provide adequate support to the patient and are extremely costly to the system. As a result, the Centers for Medicare and Medicaid Services announced in 2012 that they would no longer reimburse hospitals for services rendered if a patient was readmitted with complications within 30 days of discharge.

Given these policy changes, being able to identify and predict those patients most at risk for costly readmissions has become a pressing priority for hospital administrators.

In this project, we shall explore how to use the techniques we have learned in order to help better manage diabetes patients who have been admitted to a hospital. Our goal is to avoid patients being readmitted within 30 days of discharge, which reduces costs for the hospital and improves outcomes for patients.

The original data is from the Center for Clinical and Translational Research at Virginia Commonwealth University. It covers data on diabetes patients across 130 U.S. hospitals from 1999 to 2008. There are over 100,000 unique hospital admissions in this dataset, from ~70,000 unique patients. The data includes demographic elements, such as age, gender, and race, as well as clinical attributes such as tests conducted, emergency/inpatient visits, etc. Refer to the original documentation for more details on the dataset. Three former students Spencer Luster, Matthew Lesser and Mridul Ganesh, brought this data set into the class and did a wonderful final project. We will use a subset processed by the group but with a somewhat different objective.

Goals of the analysis

1. Identify the factors predicting whether or not the patient will be readmitted within 30 days.
2. Propose a classification rule to predict if a patient will be readmitted within 30 days.

Characteristics of the Data Set

All observations have five things in common:

1. They are all hospital admissions
2. Each patient had some form of diabetes
3. The patient stayed for between 1 and 14 days.
4. The patient had laboratory tests performed on him/her.
5. The patient was given some form of medication during the visit.

The data was collected during a ten-year period from 1999 to 2008. There are over 100,000 unique hospital admissions in the data set, with ~70,000 unique patients.

Description of variables

The dataset used covers ~50 different variables to describe every hospital diabetes admission. In this section we give an overview and brief description of the variables in this dataset.

a) Patient identifiers:

- a. `encounter_id`: unique identifier for each admission
- b. `patient_nbr`: unique identifier for each patient

b) Patient Demographics:

`race`, `age`, `gender`, `weight` cover the basic demographic information associated with each patient. `Payer_code` is an additional variable that identifies which health insurance (Medicare /Medicaid / Commercial) the patient holds.

c) Admission and discharge details:

- a. `admission_source_id` and `admission_type_id` identify who referred the patient to the hospital (e.g. physician vs. emergency dept.) and what type of admission this was (Emergency vs. Elective vs. Urgent).
- b. `discharge_disposition_id` indicates where the patient was discharged to after treatment.

d) Patient Medical History:

- a. `num_outpatient`: number of outpatient visits by the patient in the year prior to the current encounter
- b. `num_inpatient`: number of inpatient visits by the patient in the year prior to the current encounter
- c. `num_emergency`: number of emergency visits by the patient in the year prior to the current encounter

e) Patient admission details:

- a. `medical_specialty`: the specialty of the physician admitting the patient
- b. `diag_1`, `diag_2`, `diag_3`: ICD9 codes for the primary, secondary and tertiary diagnoses of the patient. ICD9 are the universal codes that all physicians use to record diagnoses. There are various easy to use tools to lookup what individual codes mean (Wikipedia is pretty decent on its own)
- c. `time_in_hospital`: the patient's length of stay in the hospital (in days)
- d. `number_diagnoses`: Total no. of diagnosis entered for the patient
- e. `num_lab_procedures`: No. of lab procedures performed in the current encounter
- f. `num_procedures`: No. of non-lab procedures performed in the current encounter
- g. `num_medications`: No. of distinct medications prescribed in the current encounter

f) Clinical Results:

- a. `max_glu_serum`: indicates results of the glucose serum test
- b. `A1Cresult`: indicates results of the A1c test

g) Medication Details:

- a. **diabetesMed**: indicates if any diabetes medication was prescribed
- b. **change**: indicates if there was a change in diabetes medication
- c. **24 medication variables**: indicate whether the dosage of the medicines was changed in any manner during the encounter

h) Readmission indicator:

Indicates whether a patient was readmitted after a particular admission. There are 3 levels for this variable: “NO” = no readmission, “< 30” = readmission within 30 days and “> 30” = readmission after more than 30 days. The 30 day distinction is of practical importance to hospitals because federal regulations penalize hospitals for an excessive proportion of such readmissions.

To save your time we are going to use some data sets cleaned by the group. Thus, we provide two datasets:

diabetic.data.csv is the original data. You may use it for the purpose of summary if you wish. You will see that the original data can’t be used directly for your analysis, yet.

readmission.csv is a cleaned version and they are modified in the following ways:

- 1) Payer code, weight and Medical Specialty are not included since they have a large number of missing values.
- 2) Variables such as **acetoexamide** (col 30), **glimepiride.pioglitazone** (45), **metformin.rosiglitazone**(46), **metformin.pioglitazone**(47) have little variability, and are as such excluded. This also includes the following variables: **chlorpropamide**(28), **acetoexamide**(30), **tolbutamide**(33), **acarbose**(36), **miglitor**(37), **troglitazone**(38), **tolazamide**(39), **examide**(40), **citoglipton**(41), **glyburide.metformin**(43), **glipizide.metformin**(44), and **glimepiride.pioglitazone**(45).
- 3) Some categorical variables have been regrouped. For example, **Diag1_mod** keeps some original levels with large number of patients and aggregates other patients as **others**. This process is known as ‘binning.’
- 4) The event of interest is **readmitted within < 30 days**. Note that you need to create this response first by regrouping **Readmission indicator**!

Exploratory Data Analysis

```
## ===== STANDARD EDA TECHNIQUES
## =====

## <<<< READING IN DATA >>>> ===== FULL DATASET ===== bill.data.test <-
## read.csv('Bills.subset.test.csv', header=TRUE, sep=',', na.strings='') #
## accounts for header, CSV, and na strings
df.full <- read.csv("diabetic.data.csv", header = TRUE, sep = ",", na.strings = "") # accounts for head
dim(df.full) # 101,766 observations x 50 variables

## [1] 101766      50

head(df.full, 30)
```

```
##   encounter_id patient_nbr      race gender      age weight
## 1      2278392     8222157   Caucasian Female  [0-10)      ?
## 2      149190     55629189   Caucasian Female  [10-20)     ?
## 3       64410     86047875 AfricanAmerican Female  [20-30)     ?
## 4      500364     82442376   Caucasian   Male   [30-40)     ?
## 5       16680     42519267   Caucasian   Male   [40-50)     ?
## 6       35754     82637451   Caucasian   Male   [50-60)     ?
## 7       55842     84259809   Caucasian   Male   [60-70)     ?
```

## 8	63768	114882984	Caucasian	Male	[70-80)	?
## 9	12522	48330783	Caucasian	Female	[80-90)	?
## 10	15738	63555939	Caucasian	Female	[90-100)	?
## 11	28236	89869032	AfricanAmerican	Female	[40-50)	?
## 12	36900	77391171	AfricanAmerican	Male	[60-70)	?
## 13	40926	85504905	Caucasian	Female	[40-50)	?
## 14	42570	77586282	Caucasian	Male	[80-90)	?
## 15	62256	49726791	AfricanAmerican	Female	[60-70)	?
## 16	73578	86328819	AfricanAmerican	Male	[60-70)	?
## 17	77076	92519352	AfricanAmerican	Male	[50-60)	?
## 18	84222	108662661	Caucasian	Female	[50-60)	?
## 19	89682	107389323	AfricanAmerican	Male	[70-80)	?
## 20	148530	69422211	?	Male	[70-80)	?
## 21	150006	22864131	?	Female	[50-60)	?
## 22	150048	21239181	?	Male	[60-70)	?
## 23	182796	63000108	AfricanAmerican	Female	[70-80)	?
## 24	183930	107400762	Caucasian	Female	[80-90)	?
## 25	216156	62718876	AfricanAmerican	Female	[70-80)	?
## 26	221634	21861756	Other	Female	[50-60)	?
## 27	236316	40523301	Caucasian	Male	[80-90)	?
## 28	248916	115196778	Caucasian	Female	[50-60)	?
## 29	250872	41606064	Caucasian	Male	[20-30)	?
## 30	252822	18196434	Caucasian	Female	[80-90)	?
##	admission_type_id	discharge_disposition_id	admission_source_id			
## 1	6	25	1			
## 2	1	1	7			
## 3	1	1	7			
## 4	1	1	7			
## 5	1	1	7			
## 6	2	1	2			
## 7	3	1	2			
## 8	1	1	7			
## 9	2	1	4			
## 10	3	3	4			
## 11	1	1	7			
## 12	2	1	4			
## 13	1	3	7			
## 14	1	6	7			
## 15	3	1	2			
## 16	1	3	7			
## 17	1	1	7			
## 18	1	1	7			
## 19	1	1	7			
## 20	3	6	2			
## 21	2	1	4			
## 22	2	1	4			
## 23	2	1	4			
## 24	2	6	1			
## 25	3	1	2			
## 26	1	1	7			
## 27	1	3	7			
## 28	1	1	1			
## 29	2	1	2			
## 30	1	2	7			

##	time_in_hospital	payer_code	medical_specialty	num_lab_procedures
## 1	1	?	Pediatrics-Endocrinology	41
## 2	3	?	?	59
## 3	2	?	?	11
## 4	2	?	?	44
## 5	1	?	?	51
## 6	3	?	?	31
## 7	4	?	?	70
## 8	5	?	?	73
## 9	13	?	?	68
## 10	12	?	InternalMedicine	33
## 11	9	?	?	47
## 12	7	?	?	62
## 13	7	?	Family/GeneralPractice	60
## 14	10	?	Family/GeneralPractice	55
## 15	1	?	?	49
## 16	12	?	?	75
## 17	4	?	?	45
## 18	3	?	Cardiology	29
## 19	5	?	?	35
## 20	6	?	?	42
## 21	2	?	?	66
## 22	2	?	?	36
## 23	2	?	?	47
## 24	11	?	?	42
## 25	3	?	?	19
## 26	1	?	?	33
## 27	6	?	Cardiology	64
## 28	2	?	Surgery-General	25
## 29	10	?	?	53
## 30	5	?	Cardiology	52
##	num_procedures	num_medications	number_outpatient	number_emergency
## 1	0	1	0	0
## 2	0	18	0	0
## 3	5	13	2	0
## 4	1	16	0	0
## 5	0	8	0	0
## 6	6	16	0	0
## 7	1	21	0	0
## 8	0	12	0	0
## 9	2	28	0	0
## 10	3	18	0	0
## 11	2	17	0	0
## 12	0	11	0	0
## 13	0	15	0	1
## 14	1	31	0	0
## 15	5	2	0	0
## 16	5	13	0	0
## 17	4	17	0	0
## 18	0	11	0	0
## 19	5	23	0	0
## 20	2	23	0	0
## 21	1	19	0	0
## 22	2	11	0	0

## 23	0	12	0	0		
## 24	2	19	0	0		
## 25	4	18	0	0		
## 26	0	7	0	0		
## 27	3	18	0	0		
## 28	2	11	0	0		
## 29	0	20	0	0		
## 30	0	14	0	0		
##	number_inpatient	diag_1	diag_2	diag_3	number_diagnoses	max_glu_serum
## 1	0	250.83	?	?	1	None
## 2	0	276	250.01	255	9	None
## 3	1	648	250	V27	6	None
## 4	0	8	250.43	403	7	None
## 5	0	197	157	250	5	None
## 6	0	414	411	250	9	None
## 7	0	414	411	V45	7	None
## 8	0	428	492	250	8	None
## 9	0	398	427	38	8	None
## 10	0	434	198	486	8	None
## 11	0	250.7	403	996	9	None
## 12	0	157	288	197	7	None
## 13	0	428	250.43	250.6	8	None
## 14	0	428	411	427	8	None
## 15	0	518	998	627	8	None
## 16	0	999	507	996	9	None
## 17	0	410	411	414	8	None
## 18	0	682	174	250	3	None
## 19	0	402	425	416	9	None
## 20	0	737	427	714	8	None
## 21	0	410	427	428	7	None
## 22	0	572	456	427	6	None
## 23	0	410	401	582	8	None
## 24	0	V57	715	V43	8	None
## 25	0	189	496	427	6	None
## 26	0	786	401	250	3	None
## 27	0	427	428	414	7	None
## 28	0	996	585	250.01	3	None
## 29	0	277	250.02	263	6	None
## 30	0	428	410	414	8	None
##	A1Cresult	metformin	repaglinide	nateglinide	chlorpropamide	glimepiride
## 1	None	No	No	No	No	No
## 2	None	No	No	No	No	No
## 3	None	No	No	No	No	No
## 4	None	No	No	No	No	No
## 5	None	No	No	No	No	No
## 6	None	No	No	No	No	No
## 7	None	Steady	No	No	No	Steady
## 8	None	No	No	No	No	No
## 9	None	No	No	No	No	No
## 10	None	No	No	No	No	No
## 11	None	No	No	No	No	No
## 12	None	No	No	No	No	No
## 13	None	Steady	Up	No	No	No
## 14	None	No	No	No	No	No

## 15	None	No	No	No	No	No
## 16	None	No	No	No	No	No
## 17	None	No	No	No	No	No
## 18	None	No	No	No	No	No
## 19	None	No	No	No	No	No
## 20	None	No	No	No	No	No
## 21	None	No	No	No	No	No
## 22	None	Steady	No	No	No	Steady
## 23	None	No	No	No	No	No
## 24	None	No	No	No	No	No
## 25	None	No	No	No	No	No
## 26	None	Steady	No	No	No	No
## 27	>7	Steady	No	No	No	No
## 28	None	No	No	No	No	No
## 29	None	No	No	No	No	No
## 30	None	Steady	No	No	No	No
##	acetohexamide	glipizide	glyburide	tolbutamide	pioglitazone	
## 1		No	No	No	No	
## 2		No	No	No	No	
## 3		No	Steady	No	No	
## 4		No	No	No	No	
## 5		No	Steady	No	No	
## 6		No	No	No	No	
## 7		No	No	No	No	
## 8		No	No	Steady	No	
## 9		No	Steady	No	No	
## 10		No	No	No	No	
## 11		No	No	No	No	
## 12		No	No	Up	No	
## 13		No	No	No	No	
## 14		No	No	No	No	
## 15		No	No	No	No	
## 16		No	No	No	No	
## 17		No	Steady	No	No	
## 18		No	No	Steady	No	
## 19		No	No	No	No	
## 20		No	No	Down	No	
## 21		No	No	No	No	
## 22		No	No	No	No	
## 23		No	No	No	No	
## 24		No	No	No	No	
## 25		No	Steady	No	No	
## 26		No	No	No	No	
## 27		No	No	Steady	No	
## 28		No	No	No	No	
## 29		No	No	No	No	
## 30		No	No	Steady	No	
##	rosiglitazone	acarbose	miglitol	troglitazone	tolazamide	examide
## 1		No	No	No	No	No
## 2		No	No	No	No	No
## 3		No	No	No	No	No
## 4		No	No	No	No	No
## 5		No	No	No	No	No
## 6		No	No	No	No	No

## 7	No	No	No	No	No	No
## 8	No	No	No	No	No	No
## 9	No	No	No	No	No	No
## 10	Steady	No	No	No	No	No
## 11	No	No	No	No	No	No
## 12	No	No	No	No	No	No
## 13	No	No	No	No	No	No
## 14	No	No	No	No	No	No
## 15	No	No	No	No	No	No
## 16	No	No	No	No	No	No
## 17	No	No	No	No	No	No
## 18	No	No	No	No	No	No
## 19	No	No	No	No	No	No
## 20	No	No	No	No	No	No
## 21	No	No	No	No	No	No
## 22	No	No	No	No	No	No
## 23	No	No	No	No	No	No
## 24	No	No	No	No	No	No
## 25	No	No	No	No	No	No
## 26	No	No	No	No	No	No
## 27	No	No	No	No	No	No
## 28	No	No	No	No	No	No
## 29	No	No	No	No	No	No
## 30	No	No	No	No	No	No
##	citoglipton	insulin	glyburide.metformin	glipizide.metformin		
## 1	No	No		No		No
## 2	No	Up		No		No
## 3	No	No		No		No
## 4	No	Up		No		No
## 5	No	Steady		No		No
## 6	No	Steady		No		No
## 7	No	Steady		No		No
## 8	No	No		No		No
## 9	No	Steady		No		No
## 10	No	Steady		No		No
## 11	No	Steady		No		No
## 12	No	Steady		No		No
## 13	No	Down		No		No
## 14	No	Steady		No		No
## 15	No	Steady		No		No
## 16	No	Up		No		No
## 17	No	Steady		No		No
## 18	No	No		No		No
## 19	No	Steady		No		No
## 20	No	Steady		No		No
## 21	No	Down		No		No
## 22	No	Steady		No		No
## 23	No	No		No		No
## 24	No	No		No		No
## 25	No	Steady		No		No
## 26	No	No		No		No
## 27	No	No		No		No
## 28	No	Steady		No		No
## 29	No	Down		No		No

## 30	No	No	No	No
##	glimepiride.pioglitazone	metformin.rosiglitazone	metformin.pioglitazone	
## 1		No	No	No
## 2		No	No	No
## 3		No	No	No
## 4		No	No	No
## 5		No	No	No
## 6		No	No	No
## 7		No	No	No
## 8		No	No	No
## 9		No	No	No
## 10		No	No	No
## 11		No	No	No
## 12		No	No	No
## 13		No	No	No
## 14		No	No	No
## 15		No	No	No
## 16		No	No	No
## 17		No	No	No
## 18		No	No	No
## 19		No	No	No
## 20		No	No	No
## 21		No	No	No
## 22		No	No	No
## 23		No	No	No
## 24		No	No	No
## 25		No	No	No
## 26		No	No	No
## 27		No	No	No
## 28		No	No	No
## 29		No	No	No
## 30		No	No	No
##	change diabetesMed	readmitted		
## 1	No	No	NO	
## 2	Ch	Yes	>30	
## 3	No	Yes	NO	
## 4	Ch	Yes	NO	
## 5	Ch	Yes	NO	
## 6	No	Yes	>30	
## 7	Ch	Yes	NO	
## 8	No	Yes	>30	
## 9	Ch	Yes	NO	
## 10	Ch	Yes	NO	
## 11	No	Yes	>30	
## 12	Ch	Yes	<30	
## 13	Ch	Yes	<30	
## 14	No	Yes	NO	
## 15	No	Yes	>30	
## 16	Ch	Yes	NO	
## 17	Ch	Yes	<30	
## 18	No	Yes	NO	
## 19	No	Yes	>30	
## 20	Ch	Yes	NO	
## 21	Ch	Yes	NO	

```
## 22      Ch      Yes      NO
## 23      No      No      NO
## 24      No      No      >30
## 25      Ch      Yes      NO
## 26      No      Yes      NO
## 27      Ch      Yes      NO
## 28      No      Yes      >30
## 29      Ch      Yes      >30
## 30      Ch      Yes      >30
```

```
View(df.full)
```

```
## Warning: running command ''/usr/bin/otool' -L '/Library/Frameworks/
## R.framework/Resources/modules/R_de.so'' had status 1
```

```
summary(df.full)
```

```
##      encounter_id      patient_nbr      race
## Min.   : 12522      Min.   : 135      ?      : 2273
## 1st Qu.: 84961194      1st Qu.: 23413221      AfricanAmerican:19210
## Median :152388987      Median : 45505143      Asian      : 641
## Mean   :165201646      Mean   : 54330401      Caucasian  :76099
## 3rd Qu.:230270888      3rd Qu.: 87545950      Hispanic   : 2037
## Max.   :443867222      Max.   :189502619      Other      : 1506
##
##      gender      age      weight
## Female      :54708      [70-80):26068      ?      :98569
## Male      :47055      [60-70):22483      [75-100) : 1336
## Unknown/Invalid: 3      [50-60):17256      [50-75) : 897
##      [80-90):17197      [100-125): 625
##      [40-50): 9685      [125-150): 145
##      [30-40): 3775      [25-50) : 97
##      (Other): 5302      (Other) : 97
## admission_type_id discharge_disposition_id admission_source_id
## Min.   :1.000      Min.   : 1.000      Min.   : 1.000
## 1st Qu.:1.000      1st Qu.: 1.000      1st Qu.: 1.000
## Median :1.000      Median : 1.000      Median : 7.000
## Mean   :2.024      Mean   : 3.716      Mean   : 5.754
## 3rd Qu.:3.000      3rd Qu.: 4.000      3rd Qu.: 7.000
## Max.   :8.000      Max.   :28.000      Max.   :25.000
##
##      time_in_hospital      payer_code      medical_specialty
## Min.   : 1.000      ?      :40256      ?      :49949
## 1st Qu.: 2.000      MC      :32439      InternalMedicine :14635
## Median : 4.000      HM      : 6274      Emergency/Trauma : 7565
## Mean   : 4.396      SP      : 5007      Family/GeneralPractice: 7440
## 3rd Qu.: 6.000      BC      : 4655      Cardiology      : 5352
## Max.   :14.000      MD      : 3532      Surgery-General  : 3099
##      (Other): 9603      (Other)      :13726
## num_lab_procedures num_procedures num_medications number_outpatient
## Min.   : 1.0      Min.   :0.00      Min.   : 1.00      Min.   : 0.0000
## 1st Qu.: 31.0      1st Qu.:0.00      1st Qu.:10.00      1st Qu.: 0.0000
## Median : 44.0      Median :1.00      Median :15.00      Median : 0.0000
## Mean   : 43.1      Mean   :1.34      Mean   :16.02      Mean   : 0.3694
## 3rd Qu.: 57.0      3rd Qu.:2.00      3rd Qu.:20.00      3rd Qu.: 0.0000
```

```

## Max. :132.0      Max. :6.00   Max. :81.00   Max. :42.0000
##
## number_emergency number_inpatient      diag_1      diag_2
## Min. : 0.0000   Min. : 0.0000   428 : 6862   276 : 6752
## 1st Qu.: 0.0000   1st Qu.: 0.0000   414 : 6581   428 : 6662
## Median : 0.0000   Median : 0.0000   786 : 4016   250 : 6071
## Mean : 0.1978   Mean : 0.6356   410 : 3614   427 : 5036
## 3rd Qu.: 0.0000   3rd Qu.: 1.0000   486 : 3508   401 : 3736
## Max. :76.0000   Max. :21.0000   427 : 2766   496 : 3305
##
##                      (Other):74419   (Other):70204
##      diag_3      number_diagnoses max_glu_serum A1Cresult
## 250 :11555   Min. : 1.000   >200: 1485   >7 : 3812
## 401 : 8289   1st Qu.: 6.000   >300: 1264   >8 : 8216
## 276 : 5175   Median : 8.000   None:96420   None:84748
## 428 : 4577   Mean : 7.423   Norm: 2597   Norm: 4990
## 427 : 3955   3rd Qu.: 9.000
## 414 : 3664   Max. :16.000
## (Other):64551
## metformin      repaglinide      nateglinide      chlorpropamide
## Down : 575     Down : 45      Down : 11      Down : 1
## No :81778     No :100227    No :101063     No :101680
## Steady:18346   Steady: 1384   Steady: 668     Steady: 79
## Up : 1067      Up : 110      Up : 24      Up : 6
##
##
##
## glimepiride      acetohexamide      glipizide      glyburide
## Down : 194     No :101765     Down : 560     Down : 564
## No :96575     Steady: 1      No :89080     No :91116
## Steady: 4670      Steady:11356   Steady: 9274
## Up : 327          Up : 770      Up : 812
##
##
##
## tolbutamide      pioglitazone      rosiglitazone      acarbose
## No :101743     Down : 118     Down : 87     Down : 3
## Steady: 23     No :94438     No :95401     No :101458
## Steady: 6976     Steady: 6100   Steady: 295
## Up : 234      Up : 178      Up : 10
##
##
##
## miglitol      troglitazone      tolazamide      examide      citoglipton
## Down : 5      No :101763     No :101727     No:101766     No:101766
## No :101728     Steady: 3      Steady: 38
## Steady: 31          Up : 1
## Up : 2
##
##
##
## insulin      glyburide.metformin glipizide.metformin
## Down :12218     Down : 6      No :101753
## No :47383     No :101060     Steady: 13
## Steady:30849     Steady: 692

```

```
## Up      :11316   Up      :      8
##
##
##
## glimepiride.pioglitazone metformin.rosiglitazone metformin.pioglitazone
## No      :101765          No      :101764          No      :101765
## Steady:    1              Steady:    2              Steady:    1
##
##
##
##
## change      diabetesMed readmitted
## Ch:47011    No :23403    <30:11357
## No:54755    Yes:78363    >30:35545
##                                NO :54864
##
##
##
##
```

```
summary(df.full$readmitted)
```

```
##    <30    >30     NO
## 11357 35545 54864
```

```
# ===== CLEANED DATASET =====
```

```
data1 <- read.csv("diabetic.data.csv", header = TRUE, sep = ",", na.strings = "") # accounts for header
dim(data1) #101766 observations x 50 variables
```

```
## [1] 101766      50
```

```
tail(data1, 20)
```

```
##      encounter_id patient_nbr      race gender      age weight
## 101747    443797298    89955270    Caucasian Male [70-80)      ?
## 101748    443804570    33230016    Caucasian Female [70-80)      ?
## 101749    443811536    189481478    Caucasian Female [40-50)      ?
## 101750    443816024    106392411    Caucasian Female [70-80)      ?
## 101751    443824292    138784172    Caucasian Female [80-90)      ?
## 101752    443835140    175326800    Caucasian Male [70-80)      ?
## 101753    443835512    139605341      Other Female [40-50)      ?
## 101754    443841992    184875899      Other Male [40-50)      ?
## 101755    443842016    183087545    Caucasian Female [70-80)      ?
## 101756    443842022    188574944      Other Female [40-50)      ?
## 101757    443842070    140199494      Other Female [60-70)      ?
## 101758    443842136    181593374    Caucasian Female [70-80)      ?
## 101759    443842340    120975314    Caucasian Female [80-90)      ?
## 101760    443842778     86472243    Caucasian Male [80-90)      ?
## 101761    443847176    50375628 AfricanAmerican Female [60-70)      ?
## 101762    443847548    100162476 AfricanAmerican Male [70-80)      ?
## 101763    443847782    74694222 AfricanAmerican Female [80-90)      ?
## 101764    443854148    41088789    Caucasian Male [70-80)      ?
## 101765    443857166    31693671    Caucasian Female [80-90)      ?
## 101766    443867222    175429310    Caucasian Male [70-80)      ?
##      admission_type_id discharge_disposition_id admission_source_id
```

## 101747	1		1	7
## 101748	1		22	7
## 101749	1		4	7
## 101750	3		6	1
## 101751	3		1	1
## 101752	3		6	1
## 101753	3		1	1
## 101754	1		1	7
## 101755	1		1	7
## 101756	1		1	7
## 101757	1		1	7
## 101758	1		1	7
## 101759	1		1	7
## 101760	1		1	7
## 101761	1		1	7
## 101762	1		3	7
## 101763	1		4	5
## 101764	1		1	7
## 101765	2		3	7
## 101766	1		1	7
##	time_in_hospital	payer_code	medical_specialty	num_lab_procedures
## 101747	4	MC	?	2
## 101748	8	MC	InternalMedicine	51
## 101749	14	MD	?	69
## 101750	3	MC	Orthopedics	27
## 101751	3	MD	?	31
## 101752	13	MC	?	77
## 101753	3	HM	?	13
## 101754	13	?	?	51
## 101755	9	?	?	50
## 101756	14	MD	?	73
## 101757	2	MD	?	46
## 101758	5	?	?	21
## 101759	5	MC	?	76
## 101760	1	MC	?	1
## 101761	6	DM	?	45
## 101762	3	MC	?	51
## 101763	5	MC	?	33
## 101764	1	MC	?	53
## 101765	10	MC	Surgery-General	45
## 101766	6	?	?	13
##	num_procedures	num_medications	number_outpatient	number_emergency
## 101747	0	7	1	0
## 101748	6	19	0	0
## 101749	0	16	0	0
## 101750	1	29	0	1
## 101751	2	24	0	0
## 101752	6	65	0	0
## 101753	1	5	0	0
## 101754	2	13	0	0
## 101755	2	33	0	0
## 101756	6	26	0	1
## 101757	6	17	1	1
## 101758	1	16	0	0

##	101759	1		22		0		1
##	101760	0		15		3		0
##	101761	1		25		3		1
##	101762	0		16		0		0
##	101763	3		18		0		0
##	101764	0		9		1		0
##	101765	2		21		0		0
##	101766	3		3		0		0
##		number_inpatient	diag_1	diag_2	diag_3	number_diagnoses		
##	101747	0	427	427	250	5		
##	101748	0	410	311	250	9		
##	101749	0	295	305	250	5		
##	101750	0	715	401	250	9		
##	101751	0	574	574	250	9		
##	101752	0	424	429	486	16		
##	101753	0	348	784	782	8		
##	101754	0	250.8	730	731	9		
##	101755	0	574	574	250.02	9		
##	101756	0	592	599	518	9		
##	101757	1	996	585	403	9		
##	101758	1	491	518	511	9		
##	101759	0	292	8	304	9		
##	101760	0	435	784	250	7		
##	101761	2	345	438	412	9		
##	101762	0	250.13	291	458	9		
##	101763	1	560	276	787	9		
##	101764	0	38	590	296	13		
##	101765	1	996	285	998	9		
##	101766	0	530	530	787	9		
##		max_glu_serum	A1Cresult	metformin	repaglinide	nateglinide		
##	101747	None	None	No	No	No		
##	101748	None	>7	No	No	No		
##	101749	None	>7	Up	No	No		
##	101750	None	Norm	Steady	No	No		
##	101751	None	None	No	No	No		
##	101752	None	Norm	No	No	No		
##	101753	None	None	Steady	No	No		
##	101754	None	None	Steady	No	No		
##	101755	None	>7	No	No	No		
##	101756	None	>8	No	No	No		
##	101757	None	None	No	No	No		
##	101758	None	None	No	No	No		
##	101759	None	None	No	No	No		
##	101760	None	None	No	No	No		
##	101761	None	None	No	No	No		
##	101762	None	>8	Steady	No	No		
##	101763	None	None	No	No	No		
##	101764	None	None	Steady	No	No		
##	101765	None	None	No	No	No		
##	101766	None	None	No	No	No		
##		chlorpropamide	glimepiride	acetoexamide	glipizide	glyburide		
##	101747	No	No	No	Steady	No		
##	101748	No	No	No	No	No		
##	101749	No	No	No	No	Steady		

## 101750	No	No	No	Steady	No
## 101751	No	No	No	No	No
## 101752	No	No	No	No	No
## 101753	No	No	No	No	Steady
## 101754	No	No	No	No	No
## 101755	No	No	No	No	Up
## 101756	No	No	No	Steady	No
## 101757	No	No	No	No	No
## 101758	No	No	No	No	No
## 101759	No	No	No	No	No
## 101760	No	No	No	No	No
## 101761	No	No	No	No	No
## 101762	No	No	No	No	No
## 101763	No	No	No	No	No
## 101764	No	No	No	No	No
## 101765	No	No	No	Steady	No
## 101766	No	No	No	No	No
##	tolbutamide	pioglitazone	rosiglitazone	acarbose	miglitol
## 101747	No	No	No	No	No
## 101748	No	No	No	No	No
## 101749	No	No	No	No	No
## 101750	No	No	No	No	No
## 101751	No	No	No	No	No
## 101752	No	No	No	No	No
## 101753	No	No	No	No	No
## 101754	No	No	No	No	No
## 101755	No	No	No	No	No
## 101756	No	No	No	No	No
## 101757	No	No	No	No	No
## 101758	No	No	No	No	No
## 101759	No	No	No	No	No
## 101760	No	No	No	No	No
## 101761	No	No	Steady	No	No
## 101762	No	No	No	No	No
## 101763	No	No	No	No	No
## 101764	No	No	No	No	No
## 101765	No	Steady	No	No	No
## 101766	No	No	No	No	No
##	troglitazone	tolazamide	examide	citoglipton	insulin
## 101747	No	No	No	No	No
## 101748	No	No	No	No	Steady
## 101749	No	No	No	No	Down
## 101750	No	No	No	No	Steady
## 101751	No	No	No	No	Down
## 101752	No	No	No	No	Up
## 101753	No	No	No	No	Steady
## 101754	No	No	No	No	Down
## 101755	No	No	No	No	Steady
## 101756	No	No	No	No	Up
## 101757	No	No	No	No	Steady
## 101758	No	No	No	No	Steady
## 101759	No	No	No	No	Up
## 101760	No	No	No	No	Up
## 101761	No	No	No	No	Down

## 101762	No	No	No	No	Down
## 101763	No	No	No	No	Steady
## 101764	No	No	No	No	Down
## 101765	No	No	No	No	Up
## 101766	No	No	No	No	No
##	glyburide.metformin	glipizide.metformin	glimepiride.pioglitazone		
## 101747	No	No	No	No	No
## 101748	No	No	No	No	No
## 101749	No	No	No	No	No
## 101750	No	No	No	No	No
## 101751	No	No	No	No	No
## 101752	No	No	No	No	No
## 101753	No	No	No	No	No
## 101754	No	No	No	No	No
## 101755	No	No	No	No	No
## 101756	No	No	No	No	No
## 101757	No	No	No	No	No
## 101758	No	No	No	No	No
## 101759	No	No	No	No	No
## 101760	No	No	No	No	No
## 101761	No	No	No	No	No
## 101762	No	No	No	No	No
## 101763	No	No	No	No	No
## 101764	No	No	No	No	No
## 101765	No	No	No	No	No
## 101766	No	No	No	No	No
##	metformin.rosiglitazone	metformin.pioglitazone	change	diabetesMed	
## 101747	No	No	No	Yes	
## 101748	No	No	No	Yes	
## 101749	No	No	Ch	Yes	
## 101750	No	No	Ch	Yes	
## 101751	No	No	Ch	Yes	
## 101752	No	No	Ch	Yes	
## 101753	No	No	Ch	Yes	
## 101754	No	No	Ch	Yes	
## 101755	No	No	Ch	Yes	
## 101756	No	No	Ch	Yes	
## 101757	No	No	No	Yes	
## 101758	No	No	No	Yes	
## 101759	No	No	Ch	Yes	
## 101760	No	No	Ch	Yes	
## 101761	No	No	Ch	Yes	
## 101762	No	No	Ch	Yes	
## 101763	No	No	No	Yes	
## 101764	No	No	Ch	Yes	
## 101765	No	No	Ch	Yes	
## 101766	No	No	No	No	
##	readmitted				
## 101747	<30				
## 101748	>30				
## 101749	>30				
## 101750	NO				
## 101751	<30				
## 101752	NO				


```
## 101753      NO
## 101754      NO
## 101755     >30
## 101756     >30
## 101757     >30
## 101758      NO
## 101759      NO
## 101760      NO
## 101761     >30
## 101762     >30
## 101763      NO
## 101764      NO
## 101765      NO
## 101766      NO
```

```
head(data1, 20)
```

##	encounter_id	patient_nbr	race	gender	age	weight
## 1	2278392	8222157	Caucasian	Female	[0-10)	?
## 2	149190	55629189	Caucasian	Female	[10-20)	?
## 3	64410	86047875	AfricanAmerican	Female	[20-30)	?
## 4	500364	82442376	Caucasian	Male	[30-40)	?
## 5	16680	42519267	Caucasian	Male	[40-50)	?
## 6	35754	82637451	Caucasian	Male	[50-60)	?
## 7	55842	84259809	Caucasian	Male	[60-70)	?
## 8	63768	114882984	Caucasian	Male	[70-80)	?
## 9	12522	48330783	Caucasian	Female	[80-90)	?
## 10	15738	63555939	Caucasian	Female	[90-100)	?
## 11	28236	89869032	AfricanAmerican	Female	[40-50)	?
## 12	36900	77391171	AfricanAmerican	Male	[60-70)	?
## 13	40926	85504905	Caucasian	Female	[40-50)	?
## 14	42570	77586282	Caucasian	Male	[80-90)	?
## 15	62256	49726791	AfricanAmerican	Female	[60-70)	?
## 16	73578	86328819	AfricanAmerican	Male	[60-70)	?
## 17	77076	92519352	AfricanAmerican	Male	[50-60)	?
## 18	84222	108662661	Caucasian	Female	[50-60)	?
## 19	89682	107389323	AfricanAmerican	Male	[70-80)	?
## 20	148530	69422211	?	Male	[70-80)	?

##	admission_type_id	discharge_disposition_id	admission_source_id
## 1	6	25	1
## 2	1	1	7
## 3	1	1	7
## 4	1	1	7
## 5	1	1	7
## 6	2	1	2
## 7	3	1	2
## 8	1	1	7
## 9	2	1	4
## 10	3	3	4
## 11	1	1	7
## 12	2	1	4
## 13	1	3	7
## 14	1	6	7
## 15	3	1	2
## 16	1	3	7

## 17	1		1	7		
## 18	1		1	7		
## 19	1		1	7		
## 20	3		6	2		
##	time_in_hospital	payer_code	medical_specialty	num_lab_procedures		
## 1	1	? Pediatrics-Endocrinology		41		
## 2	3	? ?		59		
## 3	2	? ?		11		
## 4	2	? ?		44		
## 5	1	? ?		51		
## 6	3	? ?		31		
## 7	4	? ?		70		
## 8	5	? ?		73		
## 9	13	? ?		68		
## 10	12	? InternalMedicine		33		
## 11	9	? ?		47		
## 12	7	? ?		62		
## 13	7	? Family/GeneralPractice		60		
## 14	10	? Family/GeneralPractice		55		
## 15	1	? ?		49		
## 16	12	? ?		75		
## 17	4	? ?		45		
## 18	3	? Cardiology		29		
## 19	5	? ?		35		
## 20	6	? ?		42		
##	num_procedures	num_medications	number_outpatient	number_emergency		
## 1	0	1	0	0		
## 2	0	18	0	0		
## 3	5	13	2	0		
## 4	1	16	0	0		
## 5	0	8	0	0		
## 6	6	16	0	0		
## 7	1	21	0	0		
## 8	0	12	0	0		
## 9	2	28	0	0		
## 10	3	18	0	0		
## 11	2	17	0	0		
## 12	0	11	0	0		
## 13	0	15	0	1		
## 14	1	31	0	0		
## 15	5	2	0	0		
## 16	5	13	0	0		
## 17	4	17	0	0		
## 18	0	11	0	0		
## 19	5	23	0	0		
## 20	2	23	0	0		
##	number_inpatient	diag_1	diag_2	diag_3	number_diagnoses	max_glu_serum
## 1	0	250.83	? ?		1	None
## 2	0	276	250.01	255	9	None
## 3	1	648	250	V27	6	None
## 4	0	8	250.43	403	7	None
## 5	0	197	157	250	5	None
## 6	0	414	411	250	9	None
## 7	0	414	411	V45	7	None

## 8		0	428	492	250		8	None
## 9		0	398	427	38		8	None
## 10		0	434	198	486		8	None
## 11		0	250.7	403	996		9	None
## 12		0	157	288	197		7	None
## 13		0	428	250.43	250.6		8	None
## 14		0	428	411	427		8	None
## 15		0	518	998	627		8	None
## 16		0	999	507	996		9	None
## 17		0	410	411	414		8	None
## 18		0	682	174	250		3	None
## 19		0	402	425	416		9	None
## 20		0	737	427	714		8	None
##	A1Cresult	metformin	repaglinide	nateglinide	chlorpropamide	glimepiride		
## 1	None	No	No	No	No	No		No
## 2	None	No	No	No	No	No		No
## 3	None	No	No	No	No	No		No
## 4	None	No	No	No	No	No		No
## 5	None	No	No	No	No	No		No
## 6	None	No	No	No	No	No		No
## 7	None	Steady	No	No	No	Steady		
## 8	None	No	No	No	No	No		No
## 9	None	No	No	No	No	No		No
## 10	None	No	No	No	No	No		No
## 11	None	No	No	No	No	No		No
## 12	None	No	No	No	No	No		No
## 13	None	Steady	Up	No	No	No		No
## 14	None	No	No	No	No	No		No
## 15	None	No	No	No	No	No		No
## 16	None	No	No	No	No	No		No
## 17	None	No	No	No	No	No		No
## 18	None	No	No	No	No	No		No
## 19	None	No	No	No	No	No		No
## 20	None	No	No	No	No	No		No
##	acetoexamide	glipizide	glyburide	tolbutamide	pioglitazone			
## 1	No	No	No	No	No			
## 2	No	No	No	No	No			
## 3	No	Steady	No	No	No			
## 4	No	No	No	No	No			
## 5	No	Steady	No	No	No			
## 6	No	No	No	No	No			
## 7	No	No	No	No	No			
## 8	No	No	Steady	No	No			
## 9	No	Steady	No	No	No			
## 10	No	No	No	No	No			
## 11	No	No	No	No	No			
## 12	No	No	Up	No	No			
## 13	No	No	No	No	No			
## 14	No	No	No	No	No			
## 15	No	No	No	No	No			
## 16	No	No	No	No	No			
## 17	No	Steady	No	No	No			
## 18	No	No	Steady	No	No			
## 19	No	No	No	No	No			

## 20	No	No	Down	No	No	
##	rosiglitazone	acarbose	miglitol	troglitazone	tolazamide	examide
## 1	No	No	No	No	No	No
## 2	No	No	No	No	No	No
## 3	No	No	No	No	No	No
## 4	No	No	No	No	No	No
## 5	No	No	No	No	No	No
## 6	No	No	No	No	No	No
## 7	No	No	No	No	No	No
## 8	No	No	No	No	No	No
## 9	No	No	No	No	No	No
## 10	Steady	No	No	No	No	No
## 11	No	No	No	No	No	No
## 12	No	No	No	No	No	No
## 13	No	No	No	No	No	No
## 14	No	No	No	No	No	No
## 15	No	No	No	No	No	No
## 16	No	No	No	No	No	No
## 17	No	No	No	No	No	No
## 18	No	No	No	No	No	No
## 19	No	No	No	No	No	No
## 20	No	No	No	No	No	No
##	citoglipton	insulin	glyburide.metformin	glipizide.metformin		
## 1	No	No		No		No
## 2	No	Up		No		No
## 3	No	No		No		No
## 4	No	Up		No		No
## 5	No	Steady		No		No
## 6	No	Steady		No		No
## 7	No	Steady		No		No
## 8	No	No		No		No
## 9	No	Steady		No		No
## 10	No	Steady		No		No
## 11	No	Steady		No		No
## 12	No	Steady		No		No
## 13	No	Down		No		No
## 14	No	Steady		No		No
## 15	No	Steady		No		No
## 16	No	Up		No		No
## 17	No	Steady		No		No
## 18	No	No		No		No
## 19	No	Steady		No		No
## 20	No	Steady		No		No
##	glimepiride.pioglitazone	metformin.rosiglitazone	metformin.pioglitazone			
## 1		No		No		No
## 2		No		No		No
## 3		No		No		No
## 4		No		No		No
## 5		No		No		No
## 6		No		No		No
## 7		No		No		No
## 8		No		No		No
## 9		No		No		No
## 10		No		No		No

```
## 11      No      No      No
## 12      No      No      No
## 13      No      No      No
## 14      No      No      No
## 15      No      No      No
## 16      No      No      No
## 17      No      No      No
## 18      No      No      No
## 19      No      No      No
## 20      No      No      No
```

```
##      change diabetesMed readmitted
## 1      No      No      NO
## 2      Ch      Yes     >30
## 3      No      Yes     NO
## 4      Ch      Yes     NO
## 5      Ch      Yes     NO
## 6      No      Yes     >30
## 7      Ch      Yes     NO
## 8      No      Yes     >30
## 9      Ch      Yes     NO
## 10     Ch      Yes     NO
## 11     No      Yes     >30
## 12     Ch      Yes     <30
## 13     Ch      Yes     <30
## 14     No      Yes     NO
## 15     No      Yes     >30
## 16     Ch      Yes     NO
## 17     Ch      Yes     <30
## 18     No      Yes     NO
## 19     No      Yes     >30
## 20     Ch      Yes     NO
```

```
# View(data1)
```

```
data1 <- data1[-c(6, 11:12, 28, 30, 33, 36:41, 43:47)] # getting rid of unhelpful vars
names(data1)
```

```
## [1] "encounter_id"      "patient_nbr"
## [3] "race"              "gender"
## [5] "age"               "admission_type_id"
## [7] "discharge_disposition_id" "admission_source_id"
## [9] "time_in_hospital"  "num_lab_procedures"
## [11] "num_procedures"    "num_medications"
## [13] "number_outpatient" "number_emergency"
## [15] "number_inpatient"  "diag_1"
## [17] "diag_2"            "diag_3"
## [19] "number_diagnoses"  "max_glu_serum"
## [21] "A1Cresult"         "metformin"
## [23] "repaglinide"       "nateglinide"
## [25] "glimepiride"       "glipizide"
## [27] "glyburide"         "pioglitazone"
## [29] "rosiglitazone"     "insulin"
## [31] "change"            "diabetesMed"
## [33] "readmitted"
```

```
dim(data1) # 101766 x 33
```

```
## [1] 101766      33
```

```
summary(data1)
```

```
##   encounter_id      patient_nbr      race
##   Min.   : 12522   Min.   : 135   ?       : 2273
##   1st Qu.: 84961194 1st Qu.: 23413221 AfricanAmerican:19210
##   Median :152388987 Median : 45505143 Asian       : 641
##   Mean   :165201646 Mean   : 54330401 Caucasian   :76099
##   3rd Qu.:230270888 3rd Qu.: 87545950 Hispanic    : 2037
##   Max.   :443867222 Max.   :189502619 Other       : 1506
##
##           gender      age      admission_type_id
##   Female      :54708   [70-80]:26068   Min.   :1.000
##   Male        :47055   [60-70]:22483   1st Qu.:1.000
##   Unknown/Invalid: 3   [50-60]:17256   Median :1.000
##                                     [80-90]:17197   Mean   :2.024
##                                     [40-50]: 9685   3rd Qu.:3.000
##                                     [30-40]: 3775   Max.   :8.000
##                                     (Other): 5302
##
##   discharge_disposition_id admission_source_id time_in_hospital
##   Min.   : 1.000   Min.   : 1.000   Min.   : 1.000
##   1st Qu.: 1.000   1st Qu.: 1.000   1st Qu.: 2.000
##   Median : 1.000   Median : 7.000   Median : 4.000
##   Mean   : 3.716   Mean   : 5.754   Mean   : 4.396
##   3rd Qu.: 4.000   3rd Qu.: 7.000   3rd Qu.: 6.000
##   Max.   :28.000   Max.   :25.000   Max.   :14.000
##
##   num_lab_procedures num_procedures num_medications number_outpatient
##   Min.   : 1.0   Min.   :0.00   Min.   : 1.00   Min.   : 0.0000
##   1st Qu.: 31.0   1st Qu.:0.00   1st Qu.:10.00   1st Qu.: 0.0000
##   Median : 44.0   Median :1.00   Median :15.00   Median : 0.0000
##   Mean   : 43.1   Mean   :1.34   Mean   :16.02   Mean   : 0.3694
##   3rd Qu.: 57.0   3rd Qu.:2.00   3rd Qu.:20.00   3rd Qu.: 0.0000
##   Max.   :132.0   Max.   :6.00   Max.   :81.00   Max.   :42.0000
##
##   number_emergency number_inpatient      diag_1      diag_2
##   Min.   : 0.0000   Min.   : 0.0000   428      : 6862   276      : 6752
##   1st Qu.: 0.0000   1st Qu.: 0.0000   414      : 6581   428      : 6662
##   Median : 0.0000   Median : 0.0000   786      : 4016   250      : 6071
##   Mean   : 0.1978   Mean   : 0.6356   410      : 3614   427      : 5036
##   3rd Qu.: 0.0000   3rd Qu.: 1.0000   486      : 3508   401      : 3736
##   Max.   :76.0000   Max.   :21.0000   427      : 2766   496      : 3305
##                                     (Other):74419   (Other):70204
##
##           diag_3      number_diagnoses max_glu_serum A1Cresult
##   250      :11555   Min.   : 1.000   >200: 1485   >7   : 3812
##   401      : 8289   1st Qu.: 6.000   >300: 1264   >8   : 8216
##   276      : 5175   Median : 8.000   None:96420   None:84748
##   428      : 4577   Mean   : 7.423   Norm: 2597   Norm: 4990
##   427      : 3955   3rd Qu.: 9.000
##   414      : 3664   Max.   :16.000
##   (Other):64551
```

```
## metformin      repaglinide      nateglinide      glimepiride
## Down : 575     Down : 45         Down : 11        Down : 194
## No :81778      No :100227        No :101063       No :96575
## Steady:18346   Steady: 1384      Steady: 668      Steady: 4670
## Up : 1067      Up : 110          Up : 24          Up : 327
##
##
##
## glipizide      glyburide      pioglitazone      rosiglitazone
## Down : 560     Down : 564        Down : 118       Down : 87
## No :89080      No :91116         No :94438        No :95401
## Steady:11356   Steady: 9274      Steady: 6976     Steady: 6100
## Up : 770       Up : 812          Up : 234         Up : 178
##
##
##
## insulin      change      diabetesMed readmitted
## Down :12218   Ch:47011    No :23403       <30:11357
## No :47383     No:54755    Yes:78363       >30:35545
## Steady:30849                                     NO :54864
## Up :11316
##
##
##
```

```
# <<<<<<<<< NA VALUES >>>>>>>>
```

```
sum(is.na(data1))
```

```
## [1] 0
```

```
# show how many NA values in each column
```

```
sapply(data1, function(x) sum(is.na(x))) # no 0 values
```

```
## encounter_id      patient_nbr      race
## 0                0                0
## gender            age      admission_type_id
## 0                0                0
## discharge_disposition_id      admission_source_id      time_in_hospital
## 0                0                0
## num_lab_procedures      num_procedures      num_medications
## 0                0                0
## number_outpatient      number_emergency      number_inpatient
## 0                0                0
## diag_1                diag_2                diag_3
## 0                0                0
## number_diagnoses      max_glu_serum      A1Cresult
## 0                0                0
## metformin            repaglinide      nateglinide
## 0                0                0
## glimepiride          glipizide      glyburide
## 0                0                0
## pioglitazone          rosiglitazone      insulin
## 0                0                0
## change              diabetesMed      readmitted
## 0                0                0
```

Variables of interest

Readmitted

```
summary(data1$readmitted)
```

```
##    <30    >30     NO
## 11357 35545 54864
# <30 >30 NO 11357 35545 54864
```

Race

```
# variables of interest
```

```
summary(data1$race) # boxplot readmit by race
```

```
##           ? AfricanAmerican      Asian      Caucasian
##           2273           19210           641           76099
##           Hispanic           Other
##           2037           1506
```

```
# filter by race (AfricanAmerican, Asian, Caucasian, Hispanic, Other) &&&
```

```
# ----- AfricanAmerican -----
```

```
readmit_less30.afamer <- filter(data1, race == "AfricanAmerican", readmitted ==
  "<30")
```

```
dim(readmit_less30.afamer) # 2155
```

```
## [1] 2155    33
```

```
readmit_more30.afamer <- filter(data1, race == "AfricanAmerican", readmitted ==
  ">30")
```

```
dim(readmit_more30.afamer) # 6634
```

```
## [1] 6634    33
```

```
readmit_none.afamer <- filter(data1, race == "AfricanAmerican", readmitted ==
  "NO")
```

```
dim(readmit_none.afamer) # 10421
```

```
## [1] 10421    33
```

```
slices.afamer <- c(2155, 6634, 10421)
```

```
lbls.afamer <- c("<30", ">30", "none")
```

```
pct.afamer <- round(slices.afamer/sum(slices.afamer) * 100)
```

```
lbls.afamer <- paste(lbls.afamer, "-(", pct.afamer, ")") # add percents to labels
```

```
lbls.afamer <- paste(lbls.afamer, "%", sep = "") # ad % to labels
```

```
# ----- ASIAN -----
```

```
readmit_less30.asian <- filter(data1, race == "Asian", readmitted == "<30")
```

```
dim(readmit_less30.asian) # 65
```

```
## [1] 65 33
```

```
readmit_more30.asian <- filter(data1, race == "Asian", readmitted == ">30")
```

```
dim(readmit_more30.asian) # 161
```

```
## [1] 161 33
```



```

readmit_none.asian <- filter(data1, race == "Asian", readmitted == "NO")
dim(readmit_none.asian) # 415

## [1] 415 33

slices.asian <- c(65, 161, 415)
lbls.asian <- c("<30", ">30", "none")
pct.asian <- round(slices.asian/sum(slices.asian) * 100)
lbls.asian <- paste(lbls.asian, "-(", pct.asian, ")") # add percents to labels
lbls.asian <- paste(lbls.asian, "%", sep = "") # ad % to labels

# ---- CAUCASIAN ----
readmit_less30.cau <- filter(data1, race == "Caucasian", readmitted == "<30")
dim(readmit_less30.cau) # 8592

## [1] 8592 33

readmit_more30.cau <- filter(data1, race == "Caucasian", readmitted == ">30")
dim(readmit_more30.cau) # 27124

## [1] 27124 33

readmit_none.cau <- filter(data1, race == "Caucasian", readmitted == "NO")
dim(readmit_none.cau) # 40383

## [1] 40383 33

slices.cau <- c(8592, 27124, 40383) #76099 total
lbls.cau <- c("<30", ">30", "none")
pct.cau <- round(slices.cau/sum(slices.cau) * 100)
lbls.cau <- paste(lbls.cau, "-(", pct.cau, ")") # add percents to labels
lbls.cau <- paste(lbls.cau, "%", sep = "") # ad % to labels

# ---- HISPANIC ----
readmit_less30.hisp <- filter(data1, race == "Hispanic", readmitted == "<30")
dim(readmit_less30.hisp) # 212

## [1] 212 33

readmit_more30.hisp <- filter(data1, race == "Hispanic", readmitted == ">30")
dim(readmit_more30.hisp) # 27124

## [1] 642 33

readmit_none.hisp <- filter(data1, race == "Hispanic", readmitted == "NO")
dim(readmit_none.hisp) # 40383

## [1] 1183 33

slices.hisp <- c(212, 642, 1183) #76099 total
lbls.hisp <- c("<30", ">30", "none")
pct.hisp <- round(slices.hisp/sum(slices.hisp) * 100)
lbls.hisp <- paste(lbls.hisp, "-(", pct.hisp, ")") # add percents to labels
lbls.hisp <- paste(lbls.hisp, "%", sep = "") # ad % to labels

# ---- OTHER ----
readmit_less30.oth <- filter(data1, race == "Other", readmitted == "<30")
dim(readmit_less30.oth) # 145

```

```
## [1] 145 33
```

```
readmit_more30.oth <- filter(data1, race == "Other", readmitted == ">30")
dim(readmit_more30.oth) # 446
```

```
## [1] 446 33
```

```
readmit_none.oth <- filter(data1, race == "Other", readmitted == "NO")
dim(readmit_none.oth) # 915
```

```
## [1] 915 33
```

```
slices.oth <- c(145, 446, 915)
lbls.oth <- c("<30", ">30", "none")
pct.oth <- round(slices.oth/sum(slices.oth) * 100)
lbls.oth <- paste(lbls.oth, "-(", pct.oth, ")") # add percents to labels
lbls.oth <- paste(lbls.oth, "%", sep = "") # ad % to labels
```

```
par(mfrow = c(2, 3))
```

```
pie(slices.afamer, labels = lbls.afamer, col = rainbow(length(lbls.afamer)),
    main = "Pie Chart of African American Readmits")
```

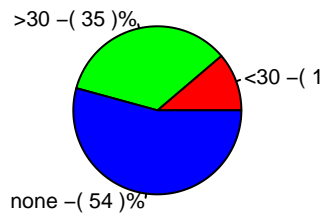
```
pie(slices.asian, labels = lbls.asian, col = rainbow(length(lbls.asian)), main = "Pie Chart of Asian Readmits")
```

```
pie(slices.cau, labels = lbls.cau, col = rainbow(length(lbls.cau)), main = "Pie Chart of Caucasian Readmits")
```

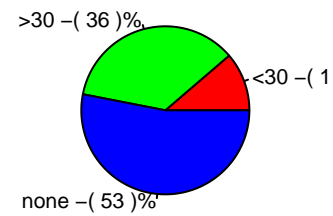
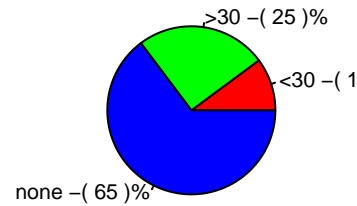
```
pie(slices.hisp, labels = lbls.hisp, col = rainbow(length(lbls.hisp)), main = "Pie Chart of Hispanic Readmits")
```

```
pie(slices.oth, labels = lbls.oth, col = rainbow(length(lbls.hisp)), main = "Pie Chart of Other Races Readmits")
```

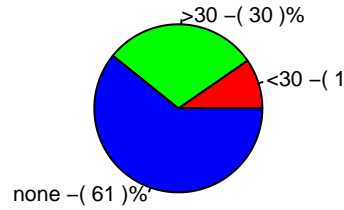
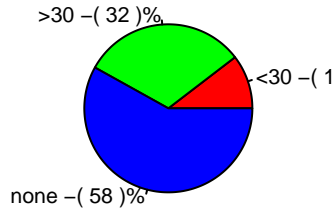
Pie Chart of African American Read



Pie Chart of Asian Readmit: Pie Chart of Caucasian Read



Pie Chart of Hispanic Readmit: Pie Chart of Other Races Read



Gender

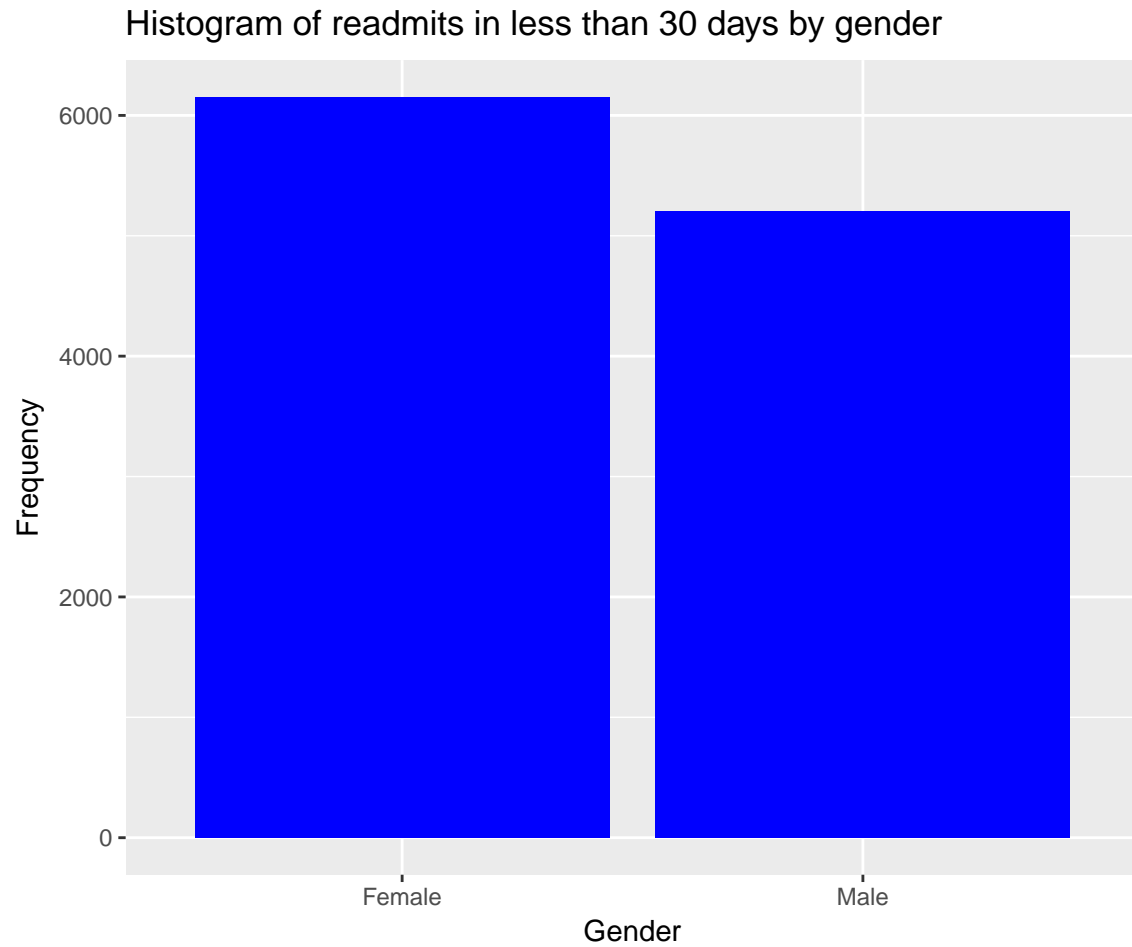
```
summary(data1$gender) #boxplot
```

```
##           Female           Male Unknown/Invalid
##           54708           47055                 3
```

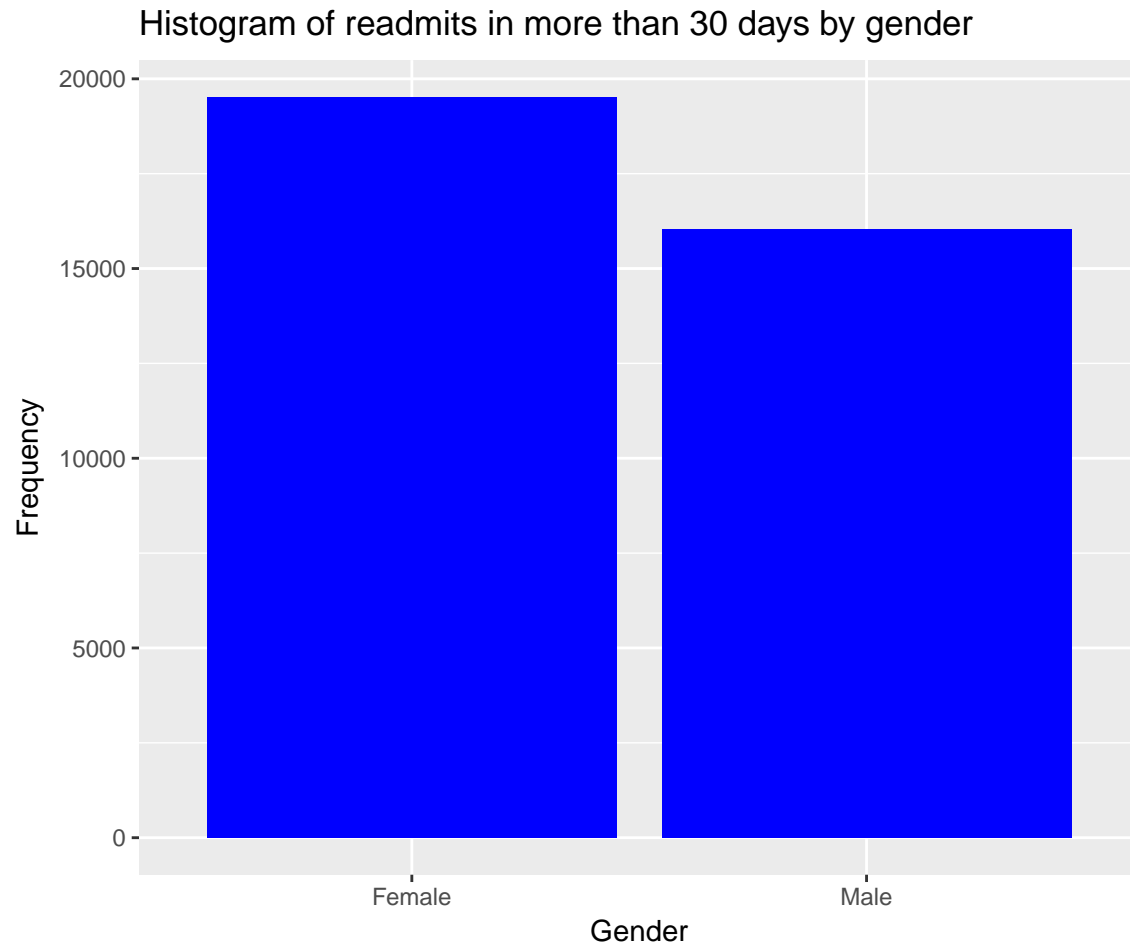
```
readmit_less30.gender <- filter(data1, readmitted == "<30")
readmit_more30.gender <- filter(data1, readmitted == ">30")
```

```
par(mfrow = c(2, 2))
```

```
ggplot(readmit_less30.gender) + geom_bar(aes(x = gender), fill = "blue") + labs(title = "Histogram of r
x = "Gender", y = "Frequency")
```



```
ggplot(readmit_more30.gender) + geom_bar(aes(x = gender), fill = "blue") + labs(title = "Histogram of r  
x = "Gender", y = "Frequency")
```



Age

```
summary(data1$age) #scatterplot
```

```
##   [0-10)  [10-20)  [20-30)  [30-40)  [40-50)  [50-60)  [60-70)  [70-80)
##      161      691     1657     3775     9685    17256    22483    26068
##  [80-90) [90-100)
##    17197     2793
```

Change (in diabetes medication)

```
summary(data1$change) #boxplot - change in diabetes medication
```

```
##   Ch   No
## 47011 54755
```

Number of diagnosis

```
summary(data1$number_diagnoses) #bar plot
```

```
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   1.000   6.000   8.000   7.423   9.000  16.000
```

```
# #<<<<< SUMMARY STATS >>>>>>> dim(county_data) #sum.isna(county_data) #
# str(county_data) summary(county_data) tail(county_data, 10)
# head(county_data, 20) View(county_data) #class(county_data)
# names(county_data) county_data[1,'adult_smoking'] # row 1 of column
# 'adult_smoking' # <<<<LEVELS of a categorical variable>>>
# levels(county_data$state) # <<<<<<setting the BASE LEVEL >>>>>> #
# levels(county_data$median_income) # 'HIGH', 'LOW', 'MEDIUM' (default is
# alphabetical order) # county_data$median_income <-
# factor(county_data$median_income, levels = c('LOW', 'MEDIUM', 'HIGH'))
# #<<<<< RENAME VARIABLES >>>>>>> #county_data <- county_data %>%
# rename(smoking_adults = adult_smoking ) # new = old name
# names(county_data) # <<<<<<<<< NA VALUES >>>>>>>>
# sum(is.na(county_data)) # show how many NA values in each column
# sapply(county_data, function(x) sum(is.na(x))) # <<<<<< REMOVE A COLUMN
# >>>>>>> #remove a column # county_data.mod <- county_data[,-1]
# #head(county_data.mod) #summary(county_data.mod) #View(county_data.mod) #
# <<<<<<<<<<< GETTING DISTINCT VALUES IN A COLUMN >>>>>>>>>>> #county_data
# %>% county_data[, -1] county_data %>% select(state) %>% distinct() # get
# distinct values for state column county_data %>% select(median_income) %>%
# distinct() # get distinct values for median_income column #county_data %>%
# select(median_income) #base R way #unique(county_data$state) # <<<<<<<
# MIN/MAX VALUES FOR A COLUMN >>>>>>>>> min(county_data$uninsured)
# max(county_data$uninsured) min(county_data$poverty_frac)
# max(county_data$poverty_frac)
```

Research approach

From the *Goals* section above, your study should respond to the following:

- 1) Identify important factors that capture the chance of a readmission within 30 days.

The set of available predictors is not limited to the raw variables in the data set. You may engineer any factors using the data, that you think will improve your model's quality.

- 2) For the purpose of classification, propose a model that can be used to predict whether a patient will be a readmit within 30 days. Justify your choice. Hint: use a decision criterion, such as AUC, to choose among a few candidate models.

Based on a quick and somewhat arbitrary guess, we estimate it costs twice as much to mislabel a readmission than it does to mislabel a non-readmission. Based on this risk ratio, propose a specific classification rule to minimize the cost. If you find any information that could provide a better cost estimate, please justify it in your write-up and use the better estimate in your answer.

Suggestion: You may use any of the methods covered so far in parts 1) and 2), and they need not be the same. Also keep in mind that a training/testing data split may be necessary.

Suggested outline

As you all know, it is very important to present your findings well. To achieve the best possible results you need to understand your audience.

Your target audience is a manager within the hospital organization. They hold an MBA, are familiar with medical terminology (though you do not need any previous medical knowledge), and have gone through a

similar course to our Modern Data Mining with someone like your professor. You can assume thus some level of technical familiarity, but should not let the paper be bogged down with code or other difficult to understand output.

Note then that the most important elements of your report are the clarity of your analysis and the quality of your proposals.

A suggested outline of the report would include the following components:

1) Executive Summary

- This section should be accessible by people with very little statistical background (avoid using technical words and no direct R output is allowed)
- Give a background of the study. You may check the original website or other sources to fill in some details, such as to why the questions we address here are important.
- A quick summary about the data.
- Methods used and the main findings.
- You may use clearly labelled and explained visualizations.
- Issues, concerns, limitations of the conclusions. This is an especially important section to be honest in - we might be Penn students, but we are statisticians today.

2) Detailed process of the analysis

i) Data Summary

- Nature of the data, origin
- Necessary quantitative and graphical summaries
- Are there any problems with the data?
- Which variables are considered as input

ii) Analyses

- Various appropriate statistical methods: e.g. glmnet
- Comparisons various models
- Final model(s)

iii) Conclusion

- Summarize results and the final model
- Final recommendations

Maintain a good descriptive flow in the text of your report. Use Appendices to display lengthy output.

iii) Appendix

- All your R code (code without comments is no good!) if you are not using `rmd` format.
- Any thing necessary to keep but for which you don't want them to be in the main report.

Collaboration

This is an **individual** assignment. We will only allow private Piazza posts for questions. If there are questions that are generally useful, we will release that information.