

# Stat 652 Project

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## Loading Libraries

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0
## v ggplot2 3.2.1      v purrr  0.3.3
## v tibble  2.1.3      v dplyr  0.8.3
## v tidyr   1.0.0      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0

## -- Conflicts ----- tidyverse_conflicts_0.3.0
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

## Loading the data

```
library(nycflights13)
library(Hmisc)
```

```
## Loading required package: lattice
## Loading required package: survival
## Loading required package: Formula
##
## Attaching package: 'Hmisc'

## The following objects are masked from 'package:dplyr':
##
##     src, summarize

## The following objects are masked from 'package:base':
##
##     format.pval, units
```

```
set.seed(42)
original_data <- read_csv("fltrain.csv.gz")
```

```
## Parsed with column specification:
## cols(
##   .default = col_double(),
##   carrier = col_character(),
##   tailnum = col_character(),
##   origin = col_character(),
##   dest = col_character(),
##   time_hour = col_datetime(format = ""),
##   name = col_character(),
##   dst = col_character(),
```

```
## tzone = col_character(),
## type = col_character(),
## manufacturer = col_character(),
## model = col_character(),
## engine = col_character()
## )

## See spec(...) for full column specifications.
DF <- original_data
```

turning all columns with datatype characters to factors.

```
DF[sapply(DF, is.character)] <- lapply(DF[sapply(DF, is.character)],
                                       as.factor)
DF$flight <- as.factor(DF$flight)
str(DF)

## Classes 'spec_tbl_df', 'tbl_df', 'tbl' and 'data.frame': 200000 obs. of  43 variables:
## $ year.x      : num  2013 2013 2013 2013 2013 ...
## $ month       : num  11 10 12 11 10 11 9 12 11 3 ...
## $ day         : num  7 30 18 20 21 7 29 21 7 31 ...
## $ dep_time    : num  600 1252 1723 2029 1620 ...
## $ sched_dep_time: num  600 1250 1715 2030 1625 ...
## $ dep_delay   : num  0 2 8 -1 -5 -8 -10 -4 0 -8 ...
## $ arr_time    : num  826 1356 2008 2141 1818 ...
## $ sched_arr_time: num  825 1400 2020 2205 1831 ...
## $ arr_delay   : num  1 -4 -12 -24 -13 -18 -10 -16 4 -11 ...
## $ carrier     : Factor w/ 16 levels "9E","AA","AS",...: 15 2 5 15 5 4 6 6 1 13 ...
## $ flight      : Factor w/ 3672 levels "1","2","3","4",...: 1525 147 1400 2343 1860 24 3083 3351 20 ...
## $ tailnum     : Factor w/ 3957 levels "D942DN","NOEGMQ",...: 1437 1226 836 565 756 2459 204 2890 6 ...
## $ origin      : Factor w/ 3 levels "EWR","JFK","LGA": 3 2 3 1 3 1 1 3 2 3 ...
## $ dest        : Factor w/ 104 levels "ABQ","ACK","ALB",...: 5 12 54 55 33 54 59 59 27 29 ...
## $ air_time    : num  123 44 133 107 90 136 110 118 101 47 ...
## $ distance    : num  762 187 950 711 502 937 725 738 589 214 ...
## $ hour        : num  6 12 17 20 16 9 15 15 16 17 ...
## $ minute      : num  0 50 15 30 25 0 29 30 50 0 ...
## $ time_hour   : POSIXct, format: "2013-11-07 11:00:00" "2013-10-30 16:00:00" ...
## $ temp        : num  63 59 34 37 63 ...
## $ dewp        : num  55.9 46.9 17.1 18 41 ...
## $ humid       : num  77.8 64.2 49.5 45.6 44.5 ...
## $ wind_dir    : num  210 240 270 20 160 240 180 190 320 140 ...
## $ wind_speed  : num  13.81 9.21 17.26 5.75 13.81 ...
## $ wind_gust   : num  NA NA 21.9 NA NA ...
## $ precip      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ pressure    : num  1011 1025 1020 1036 1017 ...
## $ visib       : num  10 10 10 10 10 10 10 10 10 10 ...
## $ name        : Factor w/ 100 levels "Akron Canton Regional Airport",...: 37 31 67 17 26 67 32 32 ...
## $ lat         : num  33.6 42.4 28.4 41.8 42.2 ...
## $ lon         : num  -84.4 -71 -81.3 -87.8 -83.4 ...
## $ alt         : num  1026 19 96 620 645 ...
## $ tz          : num  -5 -5 -5 -6 -5 -5 -6 -6 -5 -5 ...
## $ dst         : Factor w/ 2 levels "A","N": 1 1 1 1 1 1 1 1 1 1 ...
```

```

## $ tzzone      : Factor w/ 7 levels "America/Anchorage",...: 5 5 5 2 5 5 2 2 5 5 ...
## $ year.y      : num  2001 NA 2002 2006 1992 ...
## $ type        : Factor w/ 3 levels "Fixed wing multi engine",...: 1 NA 1 1 1 1 1 1 1 ...
## $ manufacturer : Factor w/ 35 levels "AGUSTA SPA","AIRBUS",...: 10 NA 2 10 3 2 18 11 11 3 ...
## $ model       : Factor w/ 126 levels "150","172E","172M",...: 37 NA 80 37 84 88 106 98 99 79 ...
## $ engines      : num   2 NA 2 2 2 2 2 2 2 ...
## $ seats        : num  140 NA 145 140 182 200 55 80 95 179 ...
## $ speed        : num   NA NA NA NA NA NA NA NA NA NA ...
## $ engine       : Factor w/ 6 levels "4 Cycle","Reciprocating",...: 3 NA 3 3 4 3 3 3 3 ...
## - attr(*, "spec")=
## .. cols(
## ..   year.x = col_double(),
## ..   month = col_double(),
## ..   day = col_double(),
## ..   dep_time = col_double(),
## ..   sched_dep_time = col_double(),
## ..   dep_delay = col_double(),
## ..   arr_time = col_double(),
## ..   sched_arr_time = col_double(),
## ..   arr_delay = col_double(),
## ..   carrier = col_character(),
## ..   flight = col_double(),
## ..   tailnum = col_character(),
## ..   origin = col_character(),
## ..   dest = col_character(),
## ..   air_time = col_double(),
## ..   distance = col_double(),
## ..   hour = col_double(),
## ..   minute = col_double(),
## ..   time_hour = col_datetime(format = ""),
## ..   temp = col_double(),
## ..   dewp = col_double(),
## ..   humid = col_double(),
## ..   wind_dir = col_double(),
## ..   wind_speed = col_double(),
## ..   wind_gust = col_double(),
## ..   precip = col_double(),
## ..   pressure = col_double(),
## ..   visib = col_double(),
## ..   name = col_character(),
## ..   lat = col_double(),
## ..   lon = col_double(),
## ..   alt = col_double(),
## ..   tz = col_double(),
## ..   dst = col_character(),
## ..   tzzone = col_character(),
## ..   year.y = col_double(),
## ..   type = col_character(),
## ..   manufacturer = col_character(),
## ..   model = col_character(),
## ..   engines = col_double(),
## ..   seats = col_double(),
## ..   speed = col_double(),
## ..   engine = col_character()

```

```
##    .. )
```

## Methods

### Preprocessing

Data preprocessing steps include the following: - Dropping columns that contain data from after the planes' departure which may leak information about the response variable dep\_delay. - Dropping columns with too many NAs. - Impute NAs for the remaining columns. - Scaling the data to work well with methods like lasso regression.

#### - Dropping columns that contain data from after the planes' departure which may leak information about the response variable dep\_delay.

dropping the columns "dep\_time", "arr\_time", "air\_time", "arr\_delay", because that leaks the response variable. dropping column "year.x" because all the values are 2013 dropping tailnum because it produces too many dummy variable columns for one hot encoding.

```
library(lubridate)
```

```
##
```

```
## Attaching package: 'lubridate'
```

```
## The following object is masked from 'package:base':
```

```
##
```

```
##      date
```

```
DF$sched_arr_time_posix <- as.POSIXct(str_pad(as.character(DF$sched_arr_time), 4, pad="0"),format="%H%M")
DF$sched_arr_time_hour <- hour(DF$sched_arr_time_posix)
DF$sched_arr_time_minute <- minute(DF$sched_arr_time_posix)
```

```
#num minute is number of minutes since start of day for scheduled arrival time
```

```
DF$sched_arr_time_num_minute <- 60*DF$sched_arr_time_hour + DF$sched_arr_time_minute
```

```
DF$sched_dep_time_posix <- as.POSIXct(str_pad(as.character(DF$sched_dep_time),4 , pad="0"),format="%H%M")
```

```
DF$sched_dep_time_hour <- hour(DF$sched_dep_time_posix)
```

```
DF$sched_dep_time_minute <- minute(DF$sched_dep_time_posix)
```

```
#num minute is number of minutes since start of day for scheduled depival time
```

```
DF$sched_dep_time_num_minute <- 60*DF$sched_dep_time_hour + DF$sched_dep_time_minute
```

```
select(original_data, time_hour, sched_dep_time, sched_arr_time, tz, tzone)
```

```
## # A tibble: 200,000 x 5
```

```
##   time_hour      sched_dep_time sched_arr_time    tz tzone
##   <dtm>          <dbl>          <dbl> <dbl> <chr>
## 1 2013-11-07 11:00:00          600           825    -5 America/New_York
## 2 2013-10-30 16:00:00         1250          1400    -5 America/New_York
## 3 2013-12-18 22:00:00         1715          2020    -5 America/New_York
## 4 2013-11-21 01:00:00         2030          2205    -6 America/Chicago
## 5 2013-10-21 20:00:00         1625          1831    -5 America/New_York
## 6 2013-11-07 14:00:00          900          1157    -5 America/New_York
## 7 2013-09-29 19:00:00         1529          1649    -6 America/Chicago
## 8 2013-12-21 20:00:00         1530          1710    -6 America/Chicago
```

```
## 9 2013-11-07 21:00:00      1650      1906      -5 America/New_York
## 10 2013-03-31 21:00:00      1700      1821      -5 America/New_York
## # ... with 199,990 more rows
```

```
select(DF, sched_arr_time, sched_arr_time_hour)
```

```
## # A tibble: 200,000 x 2
##   sched_arr_time sched_arr_time_hour
##   <dbl>          <int>
## 1         825          8
## 2        1400         14
## 3        2020         20
## 4        2205         22
## 5        1831         18
## 6        1157         11
## 7        1649         16
## 8        1710         17
## 9        1906         19
## 10       1821         18
## # ... with 199,990 more rows
```

```
DF$sched_air_time <- DF$sched_arr_time_posix - DF$sched_dep_time_posix
drops <- c('sched_arr_time_posix', 'sched_arr_time_hour', 'sched_dep_time_posix', 'sched_dep_time_hour')
DF <- DF[, !(names(DF) %in% drops)]
```

```
drops <- c("dep_time", "arr_time", "air_time", "arr_delay", "year.x", 'tailnum')
DF <- DF[, !(names(DF) %in% drops)]
```

```
DF
```

```
## # A tibble: 200,000 x 37
##   month   day dep_delay carrier flight origin dest distance temp dewp humid
##   <dbl> <dbl>   <dbl> <fct>   <fct> <fct> <fct>   <dbl> <dbl> <dbl> <dbl>
## 1    11     7         0 WN      1716 LGA    ATL     762  63.0  55.9  77.8
## 2    10    30         2 AA       178 JFK    BOS     187  59    46.9  64.2
## 3    12    18         8 DL      1585 LGA    MCO     950  34.0  17.1  49.5
## 4    11    20        -1 WN      3494 EWR    MDW     711  37.0  18.0  45.6
## 5    10    21        -5 DL      2231 LGA    DTW     502  63.0  41    44.5
## 6    11     7        -8 B6       27   EWR    MCO     937  64.4  55.4  77.3
## 7     9    29       -10 EV      4580 EWR    MKE     725  69.1  53.1  56.7
## 8    12    21        -4 EV      5207 LGA    MKE     738  57.9  46.0  64.5
## 9    11     7         0 9E      2910 JFK    CVG     589  53.6  48.2  81.9
## 10   3    31        -8 US      2183 LGA    DCA     214  51.1  36.0  56.0
## # ... with 199,990 more rows, and 26 more variables: wind_dir <dbl>,
## #   wind_speed <dbl>, wind_gust <dbl>, precip <dbl>, pressure <dbl>,
## #   visib <dbl>, name <fct>, lat <dbl>, lon <dbl>, alt <dbl>, tz <dbl>,
## #   dst <fct>, tzone <fct>, year.y <dbl>, type <fct>, manufacturer <fct>,
## #   model <fct>, engines <dbl>, seats <dbl>, speed <dbl>, engine <fct>,
## #   sched_arr_time_minute <int>, sched_arr_time_num_minute <dbl>,
## #   sched_dep_time_minute <int>, sched_dep_time_num_minute <dbl>,
## #   sched_air_time <drtm>
```

```
## Remove columns with more than 50% NA
DF <- DF[, -which(colMeans(is.na(DF)) > 0.5)]
```

```
DF$sched_air_time <- as.numeric(DF$sched_air_time)
library(imputeMissings)
```

```

##
## Attaching package: 'imputeMissings'
## The following object is masked from 'package:Hmisc':
##
##      impute
## The following object is masked from 'package:dplyr':
##
##      compute
impute_model <- imputeMissings::compute(DF, method="median/mode")
impute_model

## $month
## [1] 7
##
## $day
## [1] 16
##
## $dep_delay
## [1] -2
##
## $carrier
## [1] "UA"
##
## $flight
## [1] "15"
##
## $origin
## [1] "EWR"
##
## $dest
## [1] "ATL"
##
## $distance
## [1] 872
##
## $temp
## [1] 57.2
##
## $dewp
## [1] 42.8
##
## $humid
## [1] 57.69
##
## $wind_dir
## [1] 220
##
## $wind_speed
## [1] 10.35702
##
## $precip
## [1] 0
##

```

```

## $pressure
## [1] 1017.5
##
## $visib
## [1] 10
##
## $name
## [1] "Hartsfield Jackson Atlanta Intl"
##
## $lat
## [1] 36.09775
##
## $lon
## [1] -83.35339
##
## $alt
## [1] 433
##
## $tz
## [1] -5
##
## $dst
## [1] "A"
##
## $tzone
## [1] "America/New_York"
##
## $year.y
## [1] 2002
##
## $type
## [1] "Fixed wing multi engine"
##
## $manufacturer
## [1] "BOEING"
##
## $model
## [1] "A320-232"
##
## $engines
## [1] 2
##
## $seats
## [1] 149
##
## $engine
## [1] "Turbo-fan"
##
## $sched_arr_time_minute
## [1] 30
##
## $sched_arr_time_num_minute
## [1] 957
##

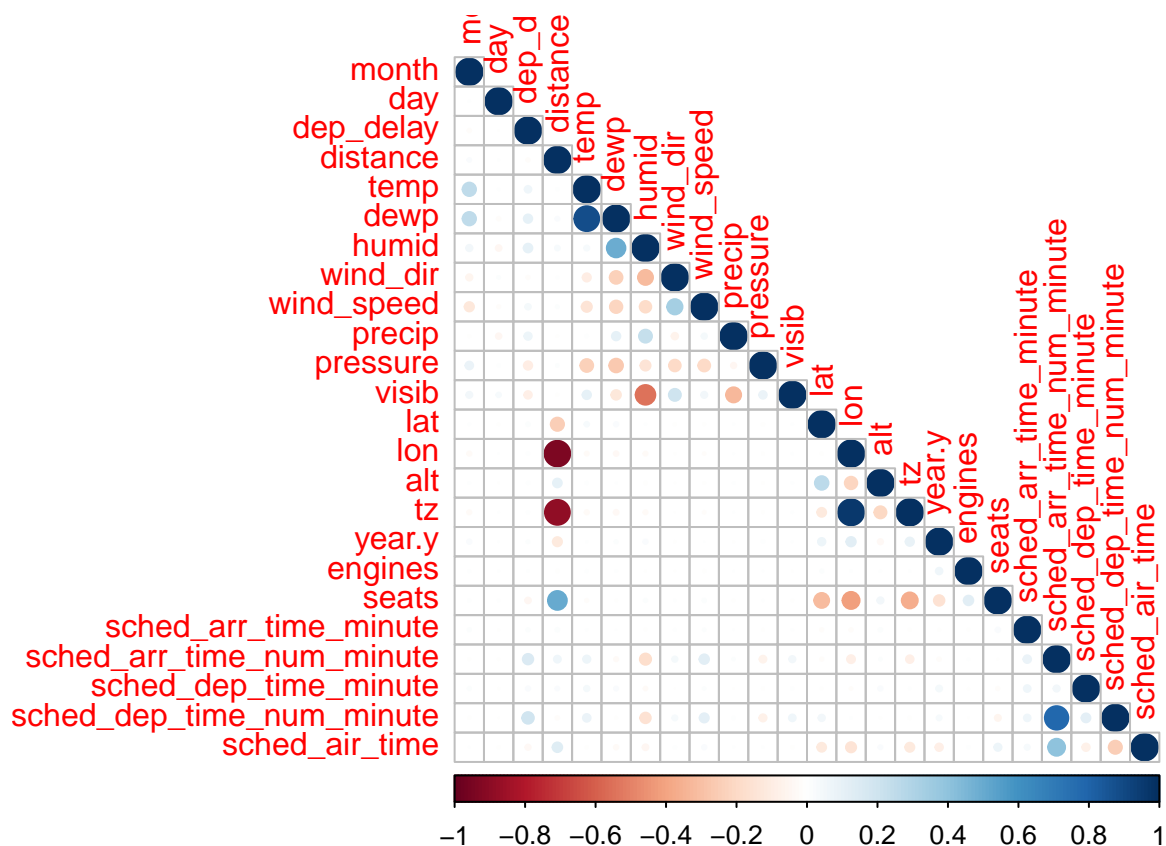
```

```
## $sched_dep_time_minute
## [1] 29
##
## $sched_dep_time_num_minute
## [1] 839
##
## $sched_air_time
## [1] 139

DF <- impute(DF, object=impute_model, , flag=TRUE)

numeric_only_df <- dplyr::select_if(DF, is.numeric)
library(corrplot)

## corrplot 0.84 loaded
corrplot(cor(numeric_only_df), type = 'lower')
```



## try features scaling

```
dep_delay_vec <- DF$dep_delay
DF$dep_delay <- NULL
head(DF)
```

```
##   month day carrier flight origin dest distance  temp  dewp  humid wind_dir
## 1    11   7      WN   1716   LGA  ATL      762 62.96 55.94 77.83      210
## 2    10  30      AA    178   JFK  BOS      187 59.00 46.94 64.22      240
```



## 3	12	18	DL	1585	LGA	MCO	950	33.98	17.06	49.51	270
## 4	11	20	WN	3494	EWR	MDW	711	37.04	17.96	45.58	20
## 5	10	21	DL	2231	LGA	DTW	502	62.96	41.00	44.47	160
## 6	11	7	B6	27	EWR	MCO	937	64.40	55.40	77.29	240
##	wind_speed	precip	pressure	visib						name	lat
## 1	13.80936	0	1011.0	10		Hartsfield Jackson Atlanta Intl				33.63672	
## 2	9.20624	0	1024.9	10		General Edward Lawrence Logan Intl				42.36435	
## 3	17.26170	0	1019.8	10		Orlando Intl				28.42939	
## 4	5.75390	0	1035.6	10		Chicago Midway Intl				41.78597	
## 5	13.80936	0	1016.9	10		Detroit Metro Wayne Co				42.21244	
## 6	16.11092	0	1017.5	10		Orlando Intl				28.42939	
##	lon	alt	tz	dst		tzone	year.y				type
## 1	-84.42807	1026	-5	A	America/New_York	2001	Fixed wing multi engine				
## 2	-71.00518	19	-5	A	America/New_York	2002	Fixed wing multi engine				
## 3	-81.30899	96	-5	A	America/New_York	2002	Fixed wing multi engine				
## 4	-87.75242	620	-6	A	America/Chicago	2006	Fixed wing multi engine				
## 5	-83.35339	645	-5	A	America/New_York	1992	Fixed wing multi engine				
## 6	-81.30899	96	-5	A	America/New_York	2006	Fixed wing multi engine				
##	manufacturer	model	engines	seats		engine	sched_arr_time_minute				
## 1	BOEING	737-7H4	2	140	Turbo-fan		25				
## 2	BOEING	A320-232	2	149	Turbo-fan		0				
## 3	AIRBUS	A319-114	2	145	Turbo-fan		20				
## 4	BOEING	737-7H4	2	140	Turbo-fan		5				
## 5	AIRBUS	INDUSTRIE A320-211	2	182	Turbo-jet		31				
## 6	AIRBUS	A320-232	2	200	Turbo-fan		57				
##	sched_arr_time_num_minute	sched_dep_time_minute	sched_dep_time_num_minute								
## 1		505	0				360				
## 2		840	50				770				
## 3		1220	15				1035				
## 4		1325	30				1230				
## 5		1111	25				985				
## 6		717	0				540				
##	sched_air_time	dep_delay_flag	temp_flag	dewp_flag	humid_flag	wind_dir_flag					
## 1	145	0	0	0	0	0					
## 2	70	0	0	0	0	0					
## 3	185	0	0	0	0	0					
## 4	95	0	0	0	0	0					
## 5	126	0	0	0	0	0					
## 6	177	0	0	0	0	0					
##	wind_speed_flag	precip_flag	pressure_flag	visib_flag	name_flag	lat_flag					
## 1	0	0	0	0	0	0					
## 2	0	0	0	0	0	0					
## 3	0	0	0	0	0	0					
## 4	0	0	0	0	0	0					
## 5	0	0	0	0	0	0					
## 6	0	0	1	0	0	0					
##	lon_flag	alt_flag	tz_flag	dst_flag	tzone_flag	year.y_flag	type_flag				
## 1	0	0	0	0	0	0	0				
## 2	0	0	0	0	0	1	1				
## 3	0	0	0	0	0	0	0				
## 4	0	0	0	0	0	0	0				
## 5	0	0	0	0	0	0	0				
## 6	0	0	0	0	0	0	0				
##	manufacturer_flag	model_flag	engines_flag	seats_flag	engine_flag						

```
## 1      0      0      0      0      0
## 2      1      1      1      1      1
## 3      0      0      0      0      0
## 4      0      0      0      0      0
## 5      0      0      0      0      0
## 6      0      0      0      0      0
```

```
library(dplyr)
DF <- DF %>% mutate_if(is.numeric, scale)
head(DF)
```

```
##      month      day carrier flight origin dest  distance      temp
## 1 1.30322 -0.9929373    WN   1716    LGA  ATL -0.3777852  0.3339858
## 2 1.01019  1.6325235    AA    178    JFK  BOS -1.1644742  0.1127815
## 3 1.59625  0.2627179    DL   1585    LGA  MCO -0.1205721 -1.2848272
## 4 1.30322  0.4910188    WN   3494    EWR  MDW -0.4475611 -1.1138966
## 5 1.01019  0.6051693    DL   2231    LGA  DTW -0.7335054  0.3339858
## 6 1.30322 -0.9929373    B6    27    EWR  MCO -0.1383581  0.4144237
##      dewp      humid      wind_dir wind_speed      precip      pressure      visib
## 1 0.7418623  0.9315583  0.07717317  0.4871566 -0.1492223 -0.96830167  0.3664282
## 2 0.2753242  0.2375566  0.36735789 -0.3415806 -0.1492223  1.01596006  0.3664282
## 3 -1.2735821 -0.5125364  0.65754261  1.1087096 -0.1492223  0.28792159  0.3664282
## 4 -1.2269283 -0.7129351 -1.76066338 -0.9631336 -0.1492223  2.54341334  0.3664282
## 5 -0.0325909 -0.7695363 -0.40646802  0.4871566 -0.1492223 -0.12606108  0.3664282
## 6 0.7138700  0.9040226  0.36735789  0.9015253 -0.1492223 -0.04040949  0.3664282
##      name      lat      lon      alt
## 1 Hartsfield Jackson Atlanta Intl -0.4207546  0.3298674  0.48364704
## 2 General Edward Lawrence Logan Intl  1.1190951  1.2378347 -0.60619417
## 3 Orlando Intl -1.3395034  0.5408516 -0.52285973
## 4 Chicago Midway Intl  1.0170501  0.1049977  0.04424731
## 5 Detroit Metro Wayne Co  1.0922942  0.4025621  0.07130394
## 6 Orlando Intl -1.3395034  0.5408516 -0.52285973
##      tz dst      tzone      year.y      type
## 1 0.6826595  A America/New_York -0.08500492 Fixed wing multi engine
## 2 0.6826595  A America/New_York  0.08617407 Fixed wing multi engine
## 3 0.6826595  A America/New_York  0.08617407 Fixed wing multi engine
## 4 -0.2514221  A America/Chicago  0.77089000 Fixed wing multi engine
## 5 0.6826595  A America/New_York -1.62561576 Fixed wing multi engine
## 6 0.6826595  A America/New_York  0.77089000 Fixed wing multi engine
##      manufacturer      model      engines      seats      engine
## 1 BOEING 737-7H4 0.05879311 0.02232546 Turbo-fan
## 2 BOEING A320-232 0.05879311 0.15869100 Turbo-fan
## 3 AIRBUS A319-114 0.05879311 0.09808410 Turbo-fan
## 4 BOEING 737-7H4 0.05879311 0.02232546 Turbo-fan
## 5 AIRBUS INDUSTRIE A320-211 0.05879311 0.65869797 Turbo-jet
## 6 AIRBUS A320-232 0.05879311 0.93142905 Turbo-fan
##      sched_arr_time_minute sched_arr_time_num_minute sched_dep_time_minute
## 1 -0.2348938 -1.4325947 -1.36042229
## 2 -1.6716145 -0.3129587  1.23408583
## 3 -0.5222379  0.9570761 -0.58206985
## 4 -1.3842703  1.3080068  0.19628258
## 5 0.1099192  0.5927766 -0.06316823
## 6 1.6041087 -0.7240489 -1.36042229
##      sched_dep_time_num_minute sched_air_time dep_delay_flag temp_flag dewp_flag
## 1 -1.6236293  0.14883297  0  0  0
```

```
## 2          -0.1673447    -0.24317221          0          0          0
## 3          0.7739125     0.35790240          0          0          0
## 4          1.4665357    -0.11250381          0          0          0
## 5          0.5963168     0.04952499          0          0          0
## 6         -0.9842849     0.31608852          0          0          0
##   humid_flag wind_dir_flag wind_speed_flag precip_flag pressure_flag visib_flag
## 1           0           0           0           0           0           0
## 2           0           0           0           0           0           0
## 3           0           0           0           0           0           0
## 4           0           0           0           0           0           0
## 5           0           0           0           0           0           0
## 6           0           0           0           0           1           0
##   name_flag lat_flag lon_flag alt_flag tz_flag dst_flag tzone_flag year.y_flag
## 1           0           0           0           0           0           0           0
## 2           0           0           0           0           0           0           1
## 3           0           0           0           0           0           0           0
## 4           0           0           0           0           0           0           0
## 5           0           0           0           0           0           0           0
## 6           0           0           0           0           0           0           0
##   type_flag manufacturer_flag model_flag engines_flag seats_flag engine_flag
## 1           0           0           0           0           0           0
## 2           1           1           1           1           1           1
## 3           0           0           0           0           0           0
## 4           0           0           0           0           0           0
## 5           0           0           0           0           0           0
## 6           0           0           0           0           0           0
```

```
DF$dep_delay <- dep_delay_vec
```

```
#take out extreme departure delays
```

```
DF<-DF[DF$dep_delay < 30,]
```

```
set.seed(42)
```

```
DF$flight <- NULL
```

```
train_index <- sample(1:nrow(DF),size=2*nrow(DF)/3,replace=FALSE)
```

```
train_df <- DF[train_index,]
```

```
test_df <- DF[-train_index,]
```

## predicting 0

```
rmse = mean((test_df$dep_delay-0)^2) %>% sqrt()
rmse
```

```
## [1] 8.30571
```

## predicting the mean

```
rmse = mean((test_df$dep_delay-mean(train_df$dep_delay))^2)%>% sqrt()
rmse
```

```
## [1] 8.299767
```

## predicting the median

```
rmse = mean((test_df$dep_delay - median(train_df$dep_delay))^2) %>% sqrt()
rmse
```

```
## [1] 8.469257
```

## linear regression

```
model <- lm(dep_delay ~ .-model, data=train_df)

summary <- round(summary(model)$coefficients, 6)
sorteddf <- summary[order(summary[, ncol(summary)]), ]
head(sorteddf)
```

```
##           Estimate Std. Error   t value Pr(>|t|)
## carrierAA -2.318607   0.269781  -8.594395     0
## carrierAS -3.341812   0.616737  -5.418535     0
## carrierDL -1.392992   0.253561  -5.493710     0
## carrierEV  1.274029   0.187960   6.778204     0
## carrierMQ -2.396562   0.272333  -8.800101     0
## carrierUS -2.323719   0.260042  -8.935921     0
```

```
sorteddf
```

##	Estimate	Std. Error	t value
## carrierAA	-2.318607	0.269781	-8.594395
## carrierAS	-3.341812	0.616737	-5.418535
## carrierDL	-1.392992	0.253561	-5.493710
## carrierEV	1.274029	0.187960	6.778204
## carrierMQ	-2.396562	0.272333	-8.800101
## carrierUS	-2.323719	0.260042	-8.935921
## carrierWN	2.796283	0.325335	8.595102
## originJFK	0.831817	0.115446	7.205266
## wind_speed	0.294774	0.027966	10.540539
## precip	0.208324	0.029304	7.109046
## pressure	-0.293555	0.027097	-10.833530
## year.y	0.257028	0.042093	6.106234
## seats	0.365535	0.068252	5.355693
## sched_arr_time_num_minute	0.241584	0.045987	5.253282
## sched_dep_time_minute	0.132111	0.024902	5.305163
## sched_dep_time_num_minute	1.258505	0.044951	27.996992
## dep_delay_flag1	-3.014132	0.155718	-19.356329
## pressure_flag1	0.606457	0.098873	6.133711
## carrierHA	-4.771275	1.108689	-4.303528
## destCHO	-36.123526	9.266129	-3.898449
## destILM	-31.097284	8.056795	-3.859759
## destPDX	14.661963	3.837985	3.820224
## destSMF	16.425104	4.321366	3.800905
## destPHX	8.055347	2.133058	3.776432
## destCRW	-31.215951	8.376979	-3.726397
## visib	-0.127430	0.034660	-3.676623
## destSAN	13.889006	3.780985	3.673383

## destJAC	11.591104	3.169410	3.657181
## destSEA	13.325506	3.646194	3.654635
## destLAX	14.348648	3.943394	3.638654
## destCAK	-31.214329	8.580927	-3.637640
## destTUL	-13.660070	3.754857	-3.637973
## destOMA	-15.118697	4.158319	-3.635771
## destSFO	16.632081	4.588360	3.624842
## destBHM	-21.009316	5.837679	-3.598916
## destLGB	14.042295	3.939582	3.564412
## destHNL	66.885768	18.771149	3.563222
## destDSM	-17.187039	4.854017	-3.540786
## destTYS	-25.105682	7.108433	-3.531817
## destBGR	-30.575425	8.693169	-3.517178
## destXNA	-14.613857	4.161947	-3.511303
## destBWI	-34.467570	9.818060	-3.510629
## destLAS	9.278585	2.648563	3.503253
## destAVL	-25.975094	7.418111	-3.501578
## destSNA	13.386324	3.822915	3.501601
## destBDL	-35.271411	10.155622	-3.473092
## destROC	-32.435956	9.347846	-3.469886
## destGSP	-25.376154	7.320086	-3.466647
## destCLE	-29.207281	8.425622	-3.466483
## destRIC	-31.792119	9.183695	-3.461800
## distance	-15.049604	4.352182	-3.457945
## destCVG	-25.692606	7.447115	-3.450008
## destBTX	-31.939098	9.300241	-3.434223
## destDAY	-26.344313	7.671046	-3.434253
## destBOS	-33.302948	9.738780	-3.419622
## destBUF	-31.162718	9.124996	-3.415094
## destSJC	15.585073	4.565151	3.413923
## destOAK	15.741470	4.621128	3.406413
## destPWM	-31.438770	9.240920	-3.402126
## destSLC	4.520401	1.329819	3.399261
## destGRR	-24.648547	7.256882	-3.396575
## destGSO	-27.791331	8.186192	-3.394903
## destIAD	-32.407962	9.550812	-3.393215
## destOKC	-10.626435	3.139739	-3.384496
## destPIT	-30.140890	8.909258	-3.383098
## destDCA	-32.545897	9.625597	-3.381182
## destMCI	-14.798414	4.379165	-3.379277
## destALB	-33.767488	9.997762	-3.377505
## destPHL	-34.872031	10.335187	-3.374108
## destCHS	-24.003318	7.115993	-3.373151
## destIND	-23.582074	6.995318	-3.371123
## destSTL	-19.054904	5.660461	-3.366317
## destPVD	-33.332083	9.908856	-3.363868
## destANC	36.554722	10.888763	3.357105
## destMHT	-32.248146	9.615996	-3.353594
## destSYR	-32.442780	9.675578	-3.353059
## destORF	-30.723323	9.167680	-3.351265
## destCMH	-27.023263	8.067612	-3.349599
## destBNA	-21.381935	6.388039	-3.347183
## destSDF	-23.471800	7.027043	-3.340210
## destRDU	-27.895202	8.360245	-3.336649

## destTVC	-23.763734	7.123077	-3.336161
## destSAV	-22.145147	6.646547	-3.331828
## destMDW	-22.017052	6.613415	-3.329150
## destACK	-32.386998	9.753827	-3.320440
## destMKE	-21.556521	6.534602	-3.298827
## destDTW	-26.134652	7.928589	-3.296255
## destCLT	-25.088232	7.674413	-3.269075
## destJAX	-19.502019	5.989172	-3.256213
## destORD	-21.244774	6.565174	-3.235981
## destMCO	-17.147920	5.306618	-3.231422
## destMSY	-12.718760	3.935982	-3.231408
## destMSN	-19.745453	6.122372	-3.225131
## destMVY	-31.958595	9.911091	-3.224528
## day	0.075847	0.023801	3.186663
## destATL	-20.349736	6.406683	-3.176329
## destRSW	-14.406148	4.545223	-3.169514
## destSRQ	-15.035777	4.752239	-3.163935
## destTPA	-15.652896	4.951828	-3.161034
## destBUR	12.507966	3.973579	3.147783
## destCAE	-22.661801	7.428543	-3.050639
## destMSP	-14.801747	4.870956	-3.038776
## destMEM	-15.840986	5.227838	-3.030122
## destFLL	-13.585753	4.566169	-2.975307
## destMIA	-13.114128	4.444244	-2.950812
## destPSE	-4.940578	1.678873	-2.942794
## destPBI	-13.976366	4.805236	-2.908570
## (Intercept)	16.474044	5.670479	2.905230
## destAUS	-5.873576	2.038240	-2.881690
## carrierB6	-0.654102	0.227007	-2.881420
## destDFW	-7.889899	2.756116	-2.862688
## destMYR	-21.926351	7.794817	-2.812940
## originLGA	-0.325794	0.115862	-2.811914
## destIAH	-7.258532	2.613026	-2.777827
## destBQN	-4.859601	1.759015	-2.762683
## destHOU	-6.906619	2.546684	-2.712005
## humid	0.417871	0.155521	2.686914
## manufacturerGULFSTREAM AEROSPACE	8.169688	3.182253	2.567265
## manufacturerCESSNA	4.567610	1.811097	2.522013
## destBZN	5.625552	2.281507	2.465718
## precip_flag1	-9.363342	3.876804	-2.415222
## temp_flag1	8.517235	3.583399	2.376859
## carrierYV	-1.410746	0.595048	-2.370812
## manufacturerLEARJET INC	9.621229	4.098480	2.347511
## destSAT	-4.082388	1.761550	-2.317497
## carrierUA	0.589784	0.254952	2.313314
## destEYW	-11.192190	4.855385	-2.305109
## carrierFL	0.948807	0.417663	2.271705
## manufacturerMCDONNELL DOUGLAS CORPORATION	7.130518	3.144929	2.267307
## carrierOO	-4.676371	2.080031	-2.248222
## month	0.055963	0.025119	2.227950
## engineReciprocating	-4.784441	2.155649	-2.219490
## engineTurbo-jet	-7.579155	3.523834	-2.150826
## manufacturerBOEING	6.655788	3.121857	2.131996
## manufacturerBARKER JACK L	4.156889	1.949770	2.131989

## manufacturerCANADAIR LTD	10.020777	4.719325	2.123350
## manufacturerDEHAVILLAND	5.277486	2.489152	2.120194
## destPSP	9.000603	4.426308	2.033434
## manufacturerMCDONNELL DOUGLAS	6.307716	3.121992	2.020414
## manufacturerPIPER	4.028235	2.033192	1.981236
## manufacturerMCDONNELL DOUGLAS AIRCRAFT CO	6.192869	3.126076	1.981036
## engineTurbo-fan	-6.970145	3.522649	-1.978666
## manufacturerPAIR MIKE E	5.954488	3.056590	1.948082
## manufacturerCIRRUS DESIGN CORP	3.649312	1.879395	1.941748
## manufacturerAIRBUS INDUSTRIE	5.928450	3.122708	1.898496
## destSTT	-2.956627	1.572582	-1.880110
## destSJU	-2.898323	1.599828	-1.811647
## manufacturerEMBRAER	5.378862	3.116909	1.725704
## destSBN	-15.570339	9.034869	-1.723361
## manufacturerAIRBUS	5.285531	3.124349	1.691722
## wind_dir_flag1	0.269529	0.159642	1.688332
## manufacturerLEBLANC GLENN T	4.512985	2.739959	1.647100
## destMTJ	5.184644	3.150881	1.645458
## manufacturerFRIEDEMANN JON	3.889750	2.392071	1.626102
## manufacturerBOMBARDIER INC	4.655030	3.120928	1.491553
## destDEN	-2.165776	1.518672	-1.426099
## manufacturerCANADAIR	4.355101	3.145879	1.384383
## manufacturerKILDALL GARY	2.862667	2.445757	1.170462
## manufacturerAMERICAN AIRCRAFT INC	3.066792	2.683506	1.142830
## manufacturerMARZ BARRY	2.829639	2.587439	1.093606
## manufacturerAVIONS MARCEL DASSAULT	6.754935	6.699339	1.008299
## manufacturerBEECH	3.431401	3.551823	0.966096
## manufacturerJOHN G HESS	-4.810782	5.388356	-0.892811
## manufacturerSIKORSKY	3.186564	3.678961	0.866159
## manufacturerBELL	2.610957	3.284993	0.794814
## type_flag1	-0.203922	0.271119	-0.752150
## manufacturerDOUGLAS	4.008806	5.632419	0.711738
## wind_dir	0.018733	0.027026	0.693150
## sched_arr_time_minute	-0.013804	0.024404	-0.565633
## carrierVX	0.181052	0.323759	0.559219
## year.y_flag1	-0.102593	0.199432	-0.514424
## manufacturerLAMBERT RICHARD	1.058252	2.476434	0.427329
## manufacturerHURLEY JAMES LARRY	-1.997386	4.928047	-0.405310
## carrierF9	0.242613	0.637308	0.380684
## destHDN	1.137227	3.190730	0.356416
## engines	-0.051607	0.146845	-0.351438
## destEGE	0.431238	1.375901	0.313422
## wind_speed_flag1	0.449983	1.446992	0.310978
## typeFixed wing single engine	0.606625	2.103014	0.288455
## dewp	0.040676	0.302037	0.134670
## manufacturerAVIAT AIRCRAFT INC	-0.501995	3.968705	-0.126488
## temp	0.023006	0.260704	0.088244
## manufacturerROBINSON HELICOPTER CO	0.248651	3.245307	0.076619
## typeRotorcraft	-0.198373	4.095669	-0.048435
##	Pr(> t )		
## carrierAA	0.000000		
## carrierAS	0.000000		
## carrierDL	0.000000		
## carrierEV	0.000000		

## carrierMQ	0.000000
## carrierUS	0.000000
## carrierWN	0.000000
## originJFK	0.000000
## wind_speed	0.000000
## precip	0.000000
## pressure	0.000000
## year.y	0.000000
## seats	0.000000
## sched_arr_time_num_minute	0.000000
## sched_dep_time_minute	0.000000
## sched_dep_time_num_minute	0.000000
## dep_delay_flag1	0.000000
## pressure_flag1	0.000000
## carrierHA	0.000017
## destCHO	0.000097
## destILM	0.000114
## destPDX	0.000133
## destSMF	0.000144
## destPHX	0.000159
## destCRW	0.000194
## visib	0.000236
## destSAN	0.000239
## destJAC	0.000255
## destSEA	0.000258
## destLAX	0.000274
## destCAK	0.000275
## destTUL	0.000275
## destOMA	0.000277
## destSFO	0.000289
## destBHM	0.000320
## destLGB	0.000365
## destHNL	0.000366
## destDSM	0.000399
## destTYS	0.000413
## destBGR	0.000436
## destXNA	0.000446
## destBWI	0.000447
## destLAS	0.000460
## destAVL	0.000463
## destSNA	0.000463
## destBDL	0.000515
## destROC	0.000521
## destGSP	0.000527
## destCLE	0.000528
## destRIC	0.000537
## distance	0.000545
## destCVG	0.000561
## destBTV	0.000594
## destDAY	0.000594
## destBOS	0.000627
## destBUF	0.000638
## destSJC	0.000641
## destOAK	0.000658



## destPWM	0.000669
## destSLC	0.000676
## destGRR	0.000683
## destGSO	0.000687
## destIAD	0.000691
## destOKC	0.000713
## destPIT	0.000717
## destDCA	0.000722
## destMCI	0.000727
## destALB	0.000732
## destPHL	0.000741
## destCHS	0.000743
## destIND	0.000749
## destSTL	0.000762
## destPVD	0.000769
## destANC	0.000788
## destMHT	0.000798
## destSYR	0.000799
## destORF	0.000805
## destCMH	0.000810
## destBNA	0.000817
## destSDF	0.000837
## destRDU	0.000848
## destTVC	0.000850
## destSAV	0.000863
## destMDW	0.000871
## destACK	0.000899
## destMKE	0.000971
## destDTW	0.000980
## destCLT	0.001079
## destJAX	0.001129
## destORD	0.001213
## destMCO	0.001232
## destMSY	0.001232
## destMSN	0.001260
## destMVY	0.001262
## day	0.001440
## destATL	0.001492
## destRSW	0.001527
## destSRQ	0.001557
## destTPA	0.001573
## destBUR	0.001646
## destCAE	0.002284
## destMSP	0.002376
## destMEM	0.002445
## destFLL	0.002928
## destMIA	0.003170
## destPSE	0.003253
## destPBI	0.003632
## (Intercept)	0.003671
## destAUS	0.003956
## carrierB6	0.003960
## destDFW	0.004201
## destMYR	0.004910

## originLGA	0.004926
## destIAH	0.005473
## destBQN	0.005734
## destHOU	0.006689
## humid	0.007213
## manufacturerGULFSTREAM AEROSPACE	0.010252
## manufacturerCESSNA	0.011670
## destBZN	0.013675
## precip_flag1	0.015727
## temp_flag1	0.017462
## carrierYV	0.017751
## manufacturerLEARJET INC	0.018901
## destSAT	0.020478
## carrierUA	0.020707
## destEYW	0.021162
## carrierFL	0.023106
## manufacturerMCDONNELL DOUGLAS CORPORATION	0.023373
## carrierOO	0.024564
## month	0.025886
## engineReciprocating	0.026455
## engineTurbo-jet	0.031492
## manufacturerBOEING	0.033009
## manufacturerBARKER JACK L	0.033010
## manufacturerCANADAIR LTD	0.033727
## manufacturerDEHAVILLAND	0.033992
## destPSP	0.042011
## manufacturerMCDONNELL DOUGLAS	0.043343
## manufacturerPIPER	0.047567
## manufacturerMCDONNELL DOUGLAS AIRCRAFT CO	0.047590
## engineTurbo-fan	0.047856
## manufacturerPAIR MIKE E	0.051408
## manufacturerCIRRUS DESIGN CORP	0.052170
## manufacturerAIRBUS INDUSTRIE	0.057633
## destSTT	0.060096
## destSJU	0.070043
## manufacturerEMBRAER	0.084403
## destSBN	0.084826
## manufacturerAIRBUS	0.090702
## wind_dir_flag1	0.091350
## manufacturerLEBLANC GLENN T	0.099540
## destMTJ	0.099878
## manufacturerFRIEDEMANN JON	0.103931
## manufacturerBOMBARDIER INC	0.135819
## destDEN	0.153843
## manufacturerCANADAIR	0.166244
## manufacturerKILDALL GARY	0.241817
## manufacturerAMERICAN AIRCRAFT INC	0.253112
## manufacturerMARZ BARRY	0.274130
## manufacturerAVIONS MARCEL DASSAULT	0.313313
## manufacturerBEECH	0.333998
## manufacturerJOHN G HESS	0.371960
## manufacturerSIKORSKY	0.386405
## manufacturerBELL	0.426724
## type_flag1	0.451962

```
## manufacturerDOUGLAS 0.476629
## wind_dir 0.488217
## sched_arr_time_minute 0.571644
## carrierVX 0.576013
## year.y_flag1 0.606957
## manufacturerLAMBERT RICHARD 0.669141
## manufacturerHURLEY JAMES LARRY 0.685251
## carrierF9 0.703438
## destHDN 0.721530
## engines 0.725260
## destEGE 0.753960
## wind_speed_flag1 0.755818
## typeFixed wing single engine 0.772999
## dewp 0.892873
## manufacturerAVIAT AIRCRAFT INC 0.899346
## temp 0.929683
## manufacturerROBINSON HELICOPTER CO 0.938927
## typeRotorcraft 0.961370

lm_test_df <- test_df

in_test_but_not_train <- setdiff(unique(lm_test_df$model), unique(train_df$model))
lm_test_df <- lm_test_df[ !lm_test_df$model %in% in_test_but_not_train, ]

in_test_but_not_train <- setdiff(unique(lm_test_df$dest), unique(train_df$dest))
lm_test_df <- lm_test_df[ !lm_test_df$dest %in% in_test_but_not_train, ]

preds = predict(model, newdata=lm_test_df)

## Warning in predict.lm(model, newdata = lm_test_df): prediction from a rank-
## deficient fit may be misleading

rmse = sqrt(mean((lm_test_df$dep_delay - preds)^2))
rmse

## [1] 8.003373
```

## gbm

```
set.seed(42)
library(gbm)

## Loaded gbm 2.1.5

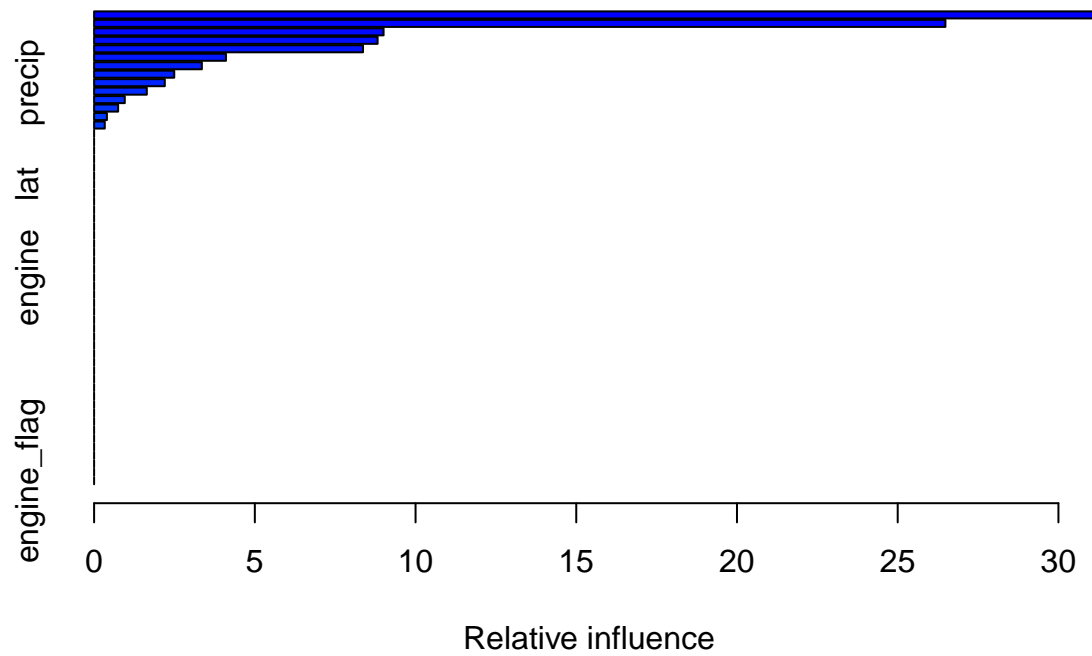
model <- gbm(dep_delay ~ ., data=train_df,
             n.trees=100, shrinkage=0.1, cv.folds = 3) # default shrinkage = 0.1

## Distribution not specified, assuming gaussian ...

preds = predict(model, newdata=test_df, n.trees=100)
rmse = sqrt(mean((test_df$dep_delay - preds)^2))
rmse

## [1] 7.978221
```

```
summary(model)
```



```
##                                var    rel.inf
## sched_dep_time_num_minute sched_dep_time_num_minute 31.1084810
## model                                model 26.4841554
## dest                                dest 9.0052224
## carrier                            carrier 8.8195760
## month                              month 8.3653954
## dewp                               dewp 4.1022280
## origin                             origin 3.3525792
## sched_arr_time_num_minute sched_arr_time_num_minute 2.4930686
## precip                             precip 2.1992862
## pressure                           pressure 1.6390248
## dep_delay_flag                     dep_delay_flag 0.9536652
## humid                              humid 0.7472539
## pressure_flag                      pressure_flag 0.3983933
## temp                               temp 0.3316705
## day                                day 0.0000000
## distance                           distance 0.0000000
## wind_dir                           wind_dir 0.0000000
## wind_speed                         wind_speed 0.0000000
## visib                              visib 0.0000000
## name                               name 0.0000000
## lat                                lat 0.0000000
## lon                                lon 0.0000000
## alt                                alt 0.0000000
## tz                                 tz 0.0000000
## dst                                dst 0.0000000
## tzone                             tzone 0.0000000
## year.y                             year.y 0.0000000
## type                               type 0.0000000
## manufacturer                       manufacturer 0.0000000
## engines                            engines 0.0000000
```

```
## seats          seats 0.0000000
## engine         engine 0.0000000
## sched_arr_time_minute sched_arr_time_minute 0.0000000
## sched_dep_time_minute sched_dep_time_minute 0.0000000
## sched_air_time   sched_air_time 0.0000000
## temp_flag        temp_flag 0.0000000
## dewp_flag        dewp_flag 0.0000000
## humid_flag       humid_flag 0.0000000
## wind_dir_flag    wind_dir_flag 0.0000000
## wind_speed_flag  wind_speed_flag 0.0000000
## precip_flag      precip_flag 0.0000000
## visib_flag       visib_flag 0.0000000
## name_flag        name_flag 0.0000000
## lat_flag         lat_flag 0.0000000
## lon_flag         lon_flag 0.0000000
## alt_flag         alt_flag 0.0000000
## tz_flag          tz_flag 0.0000000
## dst_flag         dst_flag 0.0000000
## tzone_flag       tzone_flag 0.0000000
## year.y_flag      year.y_flag 0.0000000
## type_flag        type_flag 0.0000000
## manufacturer_flag manufacturer_flag 0.0000000
## model_flag       model_flag 0.0000000
## engines_flag     engines_flag 0.0000000
## seats_flag       seats_flag 0.0000000
## engine_flag      engine_flag 0.0000000
```

Here, you can see the relative influence for each variable for gbm.

For a gbm, the improvement in the splitting criterion (which is mean squared error for regression) for a given variable is calculated at each step. The relative influence for a given variable is the average of these improvements over all the trees where the aforementioned variable is used.

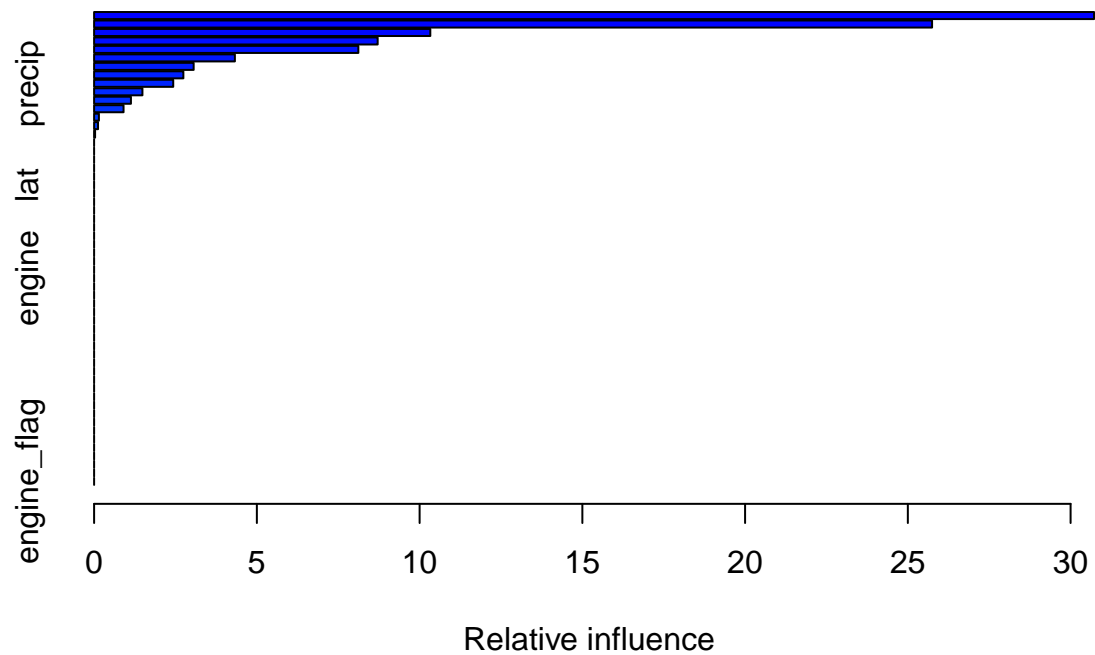
```
model <- gbm(dep_delay ~ ., data=train_df,
             n.trees=1000, shrinkage=0.01) # default shrinkage = 0.1
```

```
## Distribution not specified, assuming gaussian ...
```

```
preds = predict(model, newdata=test_df, n.trees=1000)
rmse = sqrt(mean((test_df$dep_delay - preds)^2))
rmse
```

```
## [1] 7.980782
```

```
summary(model)
```



##	var	rel.inf
## sched_dep_time_num_minute	sched_dep_time_num_minute	30.7209486
## model	model	25.7445886
## dest	dest	10.3287677
## carrier	carrier	8.7111198
## month	month	8.1184266
## dewp	dewp	4.3249508
## origin	origin	3.0586748
## sched_arr_time_num_minute	sched_arr_time_num_minute	2.7422873
## precip	precip	2.4312138
## pressure	pressure	1.4845190
## humid	humid	1.1316047
## dep_delay_flag	dep_delay_flag	0.9042069
## pressure_flag	pressure_flag	0.1476206
## temp	temp	0.1205245
## day	day	0.0305463
## distance	distance	0.0000000
## wind_dir	wind_dir	0.0000000
## wind_speed	wind_speed	0.0000000
## visib	visib	0.0000000
## name	name	0.0000000
## lat	lat	0.0000000
## lon	lon	0.0000000
## alt	alt	0.0000000
## tz	tz	0.0000000
## dst	dst	0.0000000
## tzone	tzone	0.0000000
## year.y	year.y	0.0000000
## type	type	0.0000000
## manufacturer	manufacturer	0.0000000
## engines	engines	0.0000000
## seats	seats	0.0000000
## engine	engine	0.0000000

```

## sched_arr_time_minute      sched_arr_time_minute 0.0000000
## sched_dep_time_minute     sched_dep_time_minute 0.0000000
## sched_air_time            sched_air_time 0.0000000
## temp_flag                 temp_flag 0.0000000
## dewp_flag                 dewp_flag 0.0000000
## humid_flag                humid_flag 0.0000000
## wind_dir_flag             wind_dir_flag 0.0000000
## wind_speed_flag           wind_speed_flag 0.0000000
## precip_flag               precip_flag 0.0000000
## visib_flag                visib_flag 0.0000000
## name_flag                 name_flag 0.0000000
## lat_flag                  lat_flag 0.0000000
## lon_flag                  lon_flag 0.0000000
## alt_flag                  alt_flag 0.0000000
## tz_flag                   tz_flag 0.0000000
## dst_flag                  dst_flag 0.0000000
## tzone_flag                tzone_flag 0.0000000
## year.y_flag               year.y_flag 0.0000000
## type_flag                 type_flag 0.0000000
## manufacturer_flag         manufacturer_flag 0.0000000
## model_flag                model_flag 0.0000000
## engines_flag              engines_flag 0.0000000
## seats_flag                seats_flag 0.0000000
## engine_flag               engine_flag 0.0000000

set.seed(42)

x <- 2^seq(5,14, by=1) rmse_vec <- numeric(length(x)) count <- 1 for (val in x) { hboost <- gbm( dep_delay
~ ., data = train_df, n.trees = val, distribution = 'gaussian', shrinkage = 0.01 ) preds = predict(hboost, n.trees
= val, newdata = test_df) mse = mean((test_df$dep_delay - preds) ^ 2) rmse <- sqrt(mse) rmse_vec[count]
<- rmse

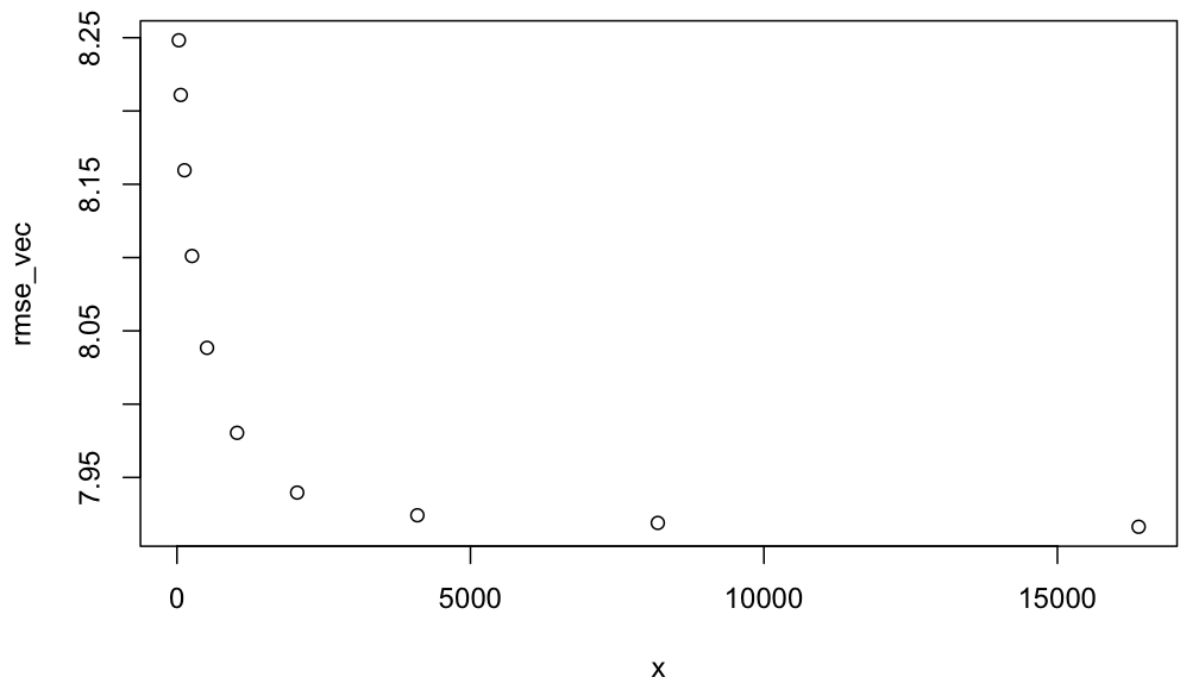
print(val) print(rmse) count = count + 1 }

plot(x, rmse_vec)

summary(hboost) class(summary(hboost)) summary <- summary(hboost) write.csv(summary,'16384trees_gbm.csv')

Analysis:

```



Tuning gbm

Here I plotted root mean squared error (rmse) vs the number of trees for shrinkage of 0.01 and all other variables as default for gbm. You can see that after around 5000 trees, increasing the number of trees further gives diminishing returns.

#Methods:

## Data Preprocessing

I performed data preprocessing. My data preprocessing steps include the following: - Dropping columns that contain data from after the planes' departure which may leak information about the response variable dep\_delay. - Dropping columns with too many NAs. - Impute NAs for the remaining columns. - Scaling the data to work well with methods like lasso regression.

## Modelling

### Basic Models

dep\_delay is the number of minutes that the plane either departs early or late. Negative numbers are for early departures and positive numbers are for the number of minutes the plane is late. First, I used a basic model of simply predicting the dep\_delay to always be 0. This was done to establish baseline performance. This model had an root mean squared error (RMSE) of



## Linear Regression

## GBM

## Basic

Conclusion: In conclusion, out of the methods that we covered in class, I found gradient boosted models to provide the best performance based on having the lowest root mean squared error on the hold out test set.

Based on the relative influence scores provided by the gbm, some of the most important feature variables include dest, model, and sched\_dep\_time\_num\_minute.

The dest column contains the airport code for where a given flight is flying to. Based on my run of gbm with a shrinkage of 0.01 and 16834 trees, dest was the most important feature with 49.56 relative influence. (“Gradient Boosting Machines · UC Business Analytics R Programming Guide” 2019).

## References

“Gradient Boosting Machines · UC Business Analytics R Programming Guide.” 2019. [http://uc-r.github.io/gbm\\_regression#h2o](http://uc-r.github.io/gbm_regression#h2o).