

# 506 Problem Set 4

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GitHub repository: [https://github.com/EmiilyLiu/STATS\\_506](https://github.com/EmiilyLiu/STATS_506)

```
setwd("F:/Desktop/STATS 506/STATS_506")
```

## Problem 1

```
## Install and load package
## install.packages("nycflights13")
#install.packages("tidyverse")

library(nycflights13)
library(tidyverse)

-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.3      v readr      2.1.4
v forcats    1.0.0      v stringr    1.5.0
v ggplot2    3.4.3      v tibble     3.2.1
v lubridate  1.9.3      v tidyr      1.3.0
v purrr      1.0.2

-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become

#data(package = "nycflights13")
```

(a)

```
## Departure table
departure_delay <- flights %>%
  group_by(origin) %>%
  summarise(
    mean_delay = mean(dep_delay, na.rm = TRUE),
    median_delay = median(dep_delay, na.rm = TRUE)
  ) %>%
  arrange(-mean_delay) %>% ## Order in descending mean delay
  ## use the airport names not the airport codes
  left_join(airports, by = c("origin" = "faa")) %>%
  select(name, mean_delay, median_delay) %>%
  print(n = Inf) ## print all rows

# A tibble: 3 x 3
```

	name	mean_delay	median_delay
	<chr>	<dbl>	<dbl>
1	Newark Liberty Intl	15.1	-1
2	John F Kennedy Intl	12.1	-1
3	La Guardia	10.3	-3

```
## Arrival table
arrival_delay <- flights %>%
  group_by(dest) %>%
  summarise(
    total_flights = n(),
    mean_delay = mean(arr_delay, na.rm = TRUE),
    median_delay = median(arr_delay, na.rm = TRUE)
  ) %>%
  filter(total_flights >= 10) %>%
  arrange(-mean_delay) %>% ## Order in descending mean delay
  ## use the airport names not the airport codes
  left_join(airports, by = c("dest" = "faa")) %>%
  select(name, mean_delay, median_delay) %>%
  ## several airports exist in table "flights" with more than 10 records
  ## but not in table "airports", ignore them
  filter(!is.na(name)) %>%
  print(n = Inf) ## print all rows
```

# A tibble: 98 x 3

	name	mean_delay	median_delay
	<chr>	<dbl>	<dbl>
1	"Columbia Metropolitan"	41.8	28
2	"Tulsa Intl"	33.7	14
3	"Will Rogers World"	30.6	16
4	"Jackson Hole Airport"	28.1	15
5	"Mc Ghee Tyson"	24.1	2
6	"Dane Co Rgnl Truax Fld"	20.2	1
7	"Richmond Intl"	20.1	1
8	"Akron Canton Regional Airport"	19.7	3
9	"Des Moines Intl"	19.0	0
10	"Gerald R Ford Intl"	18.2	1
11	"Birmingham Intl"	16.9	-2
12	"Theodore Francis Green State"	16.2	1
13	"Greenville-Spartanburg International"	15.9	-0.5
14	"Cincinnati Northern Kentucky Intl"	15.4	-3
15	"Savannah Hilton Head Intl"	15.1	-1

16	"Manchester Regional Airport"	14.8	-3
17	"Eppley Afld"	14.7	-2
18	"Yeager"	14.7	-1.5
19	"Kansas City Intl"	14.5	0
20	"Albany Intl"	14.4	-4
21	"General Mitchell Intl"	14.2	0
22	"Piedmont Triad"	14.1	-2
23	"Washington Dulles Intl"	13.9	-3
24	"Cherry Capital Airport"	13.0	-10
25	"James M Cox Dayton Intl"	12.7	-3
26	"Louisville International Airport"	12.7	-2
27	"Chicago Midway Intl"	12.4	-1
28	"Sacramento Intl"	12.1	4
29	"Jacksonville Intl"	11.8	-2
30	"Nashville Intl"	11.8	-2
31	"Portland Intl Jetport"	11.7	-4
32	"Greater Rochester Intl"	11.6	-5
33	"Hartsfield Jackson Atlanta Intl"	11.3	-1
34	"Lambert St Louis Intl"	11.1	-3
35	"Norfolk Intl"	10.9	-4
36	"Baltimore Washington Intl"	10.7	-5
37	"Memphis Intl"	10.6	-2.5
38	"Port Columbus Intl"	10.6	-3
39	"Charleston Afb Intl"	10.6	-4
40	"Philadelphia Intl"	10.1	-3
41	"Raleigh Durham Intl"	10.1	-3
42	"Indianapolis Intl"	9.94	-3
43	"Charlottesville-Albemarle"	9.5	-5
44	"Cleveland Hopkins Intl"	9.18	-5
45	"Ronald Reagan Washington Natl"	9.07	-2
46	"Burlington Intl"	8.95	-4
47	"Buffalo Niagara Intl"	8.95	-5
48	"Syracuse Hancock Intl"	8.90	-5
49	"Denver Intl"	8.61	-2
50	"Palm Beach Intl"	8.56	-3
51	"Bob Hope"	8.18	-3
52	"Fort Lauderdale Hollywood Intl"	8.08	-3
53	"Bangor Intl"	8.03	-9
54	"Asheville Regional Airport"	8.00	-1
55	"Pittsburgh Intl"	7.68	-5
56	"Gallatin Field"	7.6	-2
57	"NW Arkansas Regional"	7.47	-2
58	"Tampa Intl"	7.41	-4

59	"Charlotte Douglas Intl"	7.36	-3
60	"Minneapolis St Paul Intl"	7.27	-5
61	"William P Hobby"	7.18	-4
62	"Bradley Intl"	7.05	-10
63	"San Antonio Intl"	6.95	-9
64	"South Bend Rgnl"	6.5	-3.5
65	"Louis Armstrong New Orleans Intl"	6.49	-6
66	"Key West Intl"	6.35	7
67	"Eagle Co Rgnl"	6.30	-4
68	"Austin Bergstrom Intl"	6.02	-5
69	"Chicago Ohare Intl"	5.88	-8
70	"Orlando Intl"	5.45	-5
71	"Detroit Metro Wayne Co"	5.43	-7
72	"Portland Intl"	5.14	-5
73	"Nantucket Mem"	4.85	-3
74	"Wilmington Intl"	4.64	-7
75	"Myrtle Beach Intl"	4.60	-13
76	"Albuquerque International Sunport"	4.38	-5.5
77	"George Bush Intercontinental"	4.24	-5
78	"Norman Y Mineta San Jose Intl"	3.45	-7
79	"Southwest Florida Intl"	3.24	-5
80	"San Diego Intl"	3.14	-5
81	"Sarasota Bradenton Intl"	3.08	-5
82	"Metropolitan Oakland Intl"	3.08	-9
83	"General Edward Lawrence Logan Intl"	2.91	-9
84	"San Francisco Intl"	2.67	-8
85	"Yampa Valley"	2.14	2
86	"Phoenix Sky Harbor Intl"	2.10	-6
87	"Montrose Regional Airport"	1.79	-10.5
88	"Los Angeles Intl"	0.547	-7
89	"Dallas Fort Worth Intl"	0.322	-9
90	"Miami Intl"	0.299	-9
91	"Mc Carran Intl"	0.258	-8
92	"Salt Lake City Intl"	0.176	-8
93	"Long Beach"	-0.0620	-10
94	"Martha\\'s Vineyard"	-0.286	-11
95	"Seattle Tacoma Intl"	-1.10	-11
96	"Honolulu Intl"	-1.37	-7
97	"John Wayne Arpt Orange Co"	-7.87	-11
98	"Palm Springs Intl"	-12.7	-13.5

(b)

Although Table *planes* has a “speed” column, most of the values in this column are *NA*. Therefore, in this problem, I choose to calculate the average flights using information from Table *flights*.

```
fastest_model <- flights %>%  
  ## Calculate the average speed for each model  
  mutate(speed_mph = distance/(air_time/60)) %>%  
  inner_join(planes, by = "tailnum") %>%  
  group_by(model) %>%  
  summarise(  
    avg_speed_mph = mean(speed_mph, na.rm = TRUE),  
    total_flights = n()  
  ) %>%  
  arrange(-avg_speed_mph) %>%  
  slice(1) %>%  
  print()
```

```
# A tibble: 1 x 3  
  model   avg_speed_mph total_flights  
  <chr>         <dbl>         <int>  
1 777-222         483.             4
```

## Problem 2

```
nnmaps <- read.csv("chicago-nnmaps.csv")  
# ' get_temp - Compute the average temperature for a given month  
# ' using a specified averaging function  
# ' @param month either a numeric 1-12 or a string  
# ' @param year numeric year  
# ' @param data the data set to obtain data from  
# ' @param celsius logically indicating whether the results  
# ' should be in celsius. Default FALSE  
# ' @param average_fn a function with which to compute the mean. Default is mean  
# ' @return numeric vector of length 1 indicating  
# ' the average temperature for a given month  
get_temp <- function(month, year, data, celsius = FALSE, average_fn = mean){  
  ## Rename input parameters to avoid  
  ## potential conflicts with column names in 'data'
```

```

input_month <- month
input_year <- year

# Convert string months to numeric format
# Convert string months to numeric format
if (is.character(input_month)) {
  input_month <- case_when(
    input_month %in% c("Jan", "January") ~ 1,
    input_month %in% c("Feb", "February") ~ 2,
    input_month %in% c("Mar", "March") ~ 3,
    input_month %in% c("Apr", "April") ~ 4,
    input_month %in% c("May") ~ 5,
    input_month %in% c("Jun", "June") ~ 6,
    input_month %in% c("Jul", "July") ~ 7,
    input_month %in% c("Aug", "August") ~ 8,
    input_month %in% c("Sep", "September") ~ 9,
    input_month %in% c("Oct", "October") ~ 10,
    input_month %in% c("Nov", "November") ~ 11,
    input_month %in% c("Dec", "December") ~ 12,
    TRUE ~ -1 # Default case
  )
}

# Ensure month is between 1-12
if (input_month < 1 || input_month > 12) {
  stop("Invalid month input. Please provide a valid month")
}

# Extract the relevant data
temp_data <- data %>%
  filter(month_numeric == input_month, year == input_year) %>%
  pull(temp)

# Convert to Celsius if requested
if (celsius) {
  temp_data <- (temp_data - 32) * (5/9)
}

# Return the average temperature using the provided function
return(average_fn(temp_data))
}

```

```
get_temp("Apr", 1999, data = nnmaps)
```

```
[1] 49.8
```

```
get_temp("Apr", 1999, data = nnmaps, celsius = TRUE)
```

```
[1] 9.888889
```

```
get_temp(10, 1998, data = nnmaps, average_fn = median)
```

```
[1] 55
```

```
# get_temp(13, 1998, data = nnmaps)
# Error message for this case:
# Error in get_temp(13, 1998, data = nnmaps) :
# Invalid month input. Please provide a valid month
get_temp(2, 2005, data = nnmaps)
```

```
[1] NaN
```

```
get_temp("November", 1999, data = nnmaps, celsius = TRUE,
        average_fn = function(x) {
  x %>% sort -> x
  x[2:(length(x) - 1)] %>% mean %>% return
})
```

```
[1] 7.301587
```

### Problem 3

SAS output link: [https://github.com/EmiilyLiu/STATS\\_506](https://github.com/EmiilyLiu/STATS_506)

```
/* import csv file */
PROC IMPORT DATAFILE='/home/u63651235/sasuser.v94/recs2020_public_v5.csv'
  OUT=recs2020
```



```

                DBMS=CSV;
RUN;

/* get some information about the data */
/* I comment for the submitted version since the output is too long.
proc print data=recs2020 (obs=5);
run;

proc contents data=recs2020;
run;
*/

```

(a)

```

/* (a) */
proc sql noprint;
    /* weight of each state */
    create table state_weights as
    select state_name, sum(NWEIGHT) as total_weight
    from recs2020
    group by state_name;

    /* sum of weights of all states */
    select sum(NWEIGHT) into :totalWeight from recs2020;

    /* percentage of each state and sort */
    create table state_percent as
    select state_name,
           total_weight / &totalWeight as percentage
    from state_weights
    order by percentage desc;
run;

/* print the state with the highest percentage */
proc print data=state_percent(obs=1);
run;

/* print the percentage of Michigan */
proc print data=state_percent;
    where state_name = "Michigan";

```

```
run;
```

**(b)**

```
/* (b) */
/* get records with strictly positive value of total electricity cost */
data positive_costs;
    set recs2020;
    if DOLLAREL > 0;
run;

/* plot the histogram */
proc sgplot data=positive_costs;
    histogram DOLLAREL;
    title "Histogram of Total Electricity Cost with Strictly Positive Values";
run;
```

**(c)**

```
/* (c) */
/* calculate the log of total electricity cost */
data positive_costs;
    set positive_costs;
    log_DOLLAREL = log(DOLLAREL);
run;

/* plot the histogram */
proc sgplot data=positive_costs;
    histogram log_DOLLAREL;
    title "Histogram of Log of Total Electricity Cost";
run;
```

(d)

```
/* (d) */
/* Use the sum of NCOMBATH, NHAFBATH and TOTROOMS as the number of rooms */
data positive_costs;
    set positive_costs;
    TOTAL_ROOM = NCOMBATH + NHAFBATH + TOTROOMS;
run;

/* linear regression */
proc glm data=positive_costs;
    where PRKGPLC1 ne -2; /* Exclude observations where PRKGPLC1 is -2(missing values) */
    class PRKGPLC1;
    model log_DOLLAREL = TOTAL_ROOM PRKGPLC1; /* response and predictors */
    weight NWEIGHT; /* weights */
    title "Linear Regression on Log of Total Electricity Cost";
    output out=predicted_data p=predicted_log; /* get predicted data for (e) */
run;
```

(e)

```
/* (e) */
data predicted_data;
    set predicted_data;
    predicted_DOLLAREL = exp(predicted_log);
run;

proc sgplot data=predicted_data;
    scatter x=DOLLAREL y=predicted_DOLLAREL;
    xaxis label="Actual Total Electricity Cost";
    yaxis label="Predicted Total Electricity Cost";
    title "Scatterplot of Predicted vs Actual Total Electricity Cost";
run;
```

## Problem 4

(a)

The codebook contains the variable names, labels, and tabulations of responses for the questions asked in the survey.

(b)

```
/* (b) */
/* import csv file */
PROC IMPORT DATAFILE='/home/u63651235/sasuser.v94/public2022.csv'
      OUT=fulldata
      DBMS=CSV;
RUN;

/* get some information about the data */
proc print data=fulldata (obs=5);
run;

/* select needed variables */
proc sql;
  create table subsetData as
  select CaseID, weight_pop, B3, ND2, B7_b, GH1, race_5cat, educ_4cat
  from fullData;
run;

/* transformation */
proc format;
  value $B3fmt
    'Much worse off' = 1
    'Somewhat worse off' = 2
    'About the same' = 3
    'Somewhat better off' = 4
    'Much better off' = 5;

  value $ND2fmt
    'Much higher' = 1
    'Somewhat higher' = 2
    'About the same' = 3
    'Somewhat lower' = 4
```

```

'Much lower' = 5;

value $B7_bfmt
'Poor' = 1
'Only fair' = 2
'Good' = 3
'Excellent' = 4;

value $GH1fmt
'Own your home with a mortgage or loan' = 1
'Own your home free and clear (without a mortgage or loan)' = 2
'Pay rent' = 3
'Neither own nor pay rent' = 4;

value $race_fmt
'White' = 1
'Black' = 2
'Hispanic' = 3
'Asian' = 4
'Other' = 5;

value $edu_fmt
'Less than a high school degree' = 1
'High school degree or GED' = 2
'Some college/technical or associates degree' = 3
'Bachelor's degree or more' = 4;
run;

data Data_trans;
  set subsetData;

  B3_trans = put(B3, B3fmt.);
  ND2_trans = put(ND2, ND2fmt.);
  B7_b_trans = put(B7_b, B7_bfmt.);
  GH1_trans = put(GH1, GH1fmt.);
  race_trans = put(race_5cat, race_fmt.);
  educ_trans = put(educ_4cat, edu_fmt.);
run;

```

(c)

```
/* (c) */
/* export the transformed data into Stata format */
proc sql;
    create table outputData as
    select CaseID, weight_pop, B3_trans, ND2_trans, B7_b_trans, GH1_trans, race_trans, edu
    from Data_trans;
run;
proc export data=outputData
    outfile="/home/u63651235/sasuser.v94/public2022.dta"
    dbms=stata replace;
run;

proc print data=outputData (obs=5);
run;

proc contents data=outputData;
run;
```

(d)

```
. do "C:\Users\LOCAL_~3\Temp\STD25c4_000000.tmp"

. * (d) import data
. use "K:\public2022.dta", clear

. count
11,667

. describe
```

Contains data from K:\public2022.dta

Observations: 11,667

Variables: 8

Variable	Storage	Display	Value	
name	type	format	label	Variable label
CaseID	double	%12.0g		

```

weight_pop      double   %12.0g
B3_trans        str2     %2s
ND2_trans        str2     %2s
B7_b_trans       str2     %2s
GH1_trans        str2     %2s
race_trans       str2     %2s
educ_trans       str2     %2s

```

---

Sorted by:

There are total 11,667 observations, aligning with that in the codebook and 8 selected variables, aligning with output from SAS.

(e)

```

. * (e) transform response to a binary variable
. gen B3_bin = cond(inlist(B3_trans, "1", "2"), 0, 1)

```

(f)

```

. * (f) logisitic regression model
. svyset CaseID [pw=weight_pop]

Sampling weights: weight_pop
                  VCE: linearized
                  Single unit: missing
                  Strata 1: <one>
Sampling unit 1: CaseID
                  FPC 1: <zero>

.
. * convert string variable to numeric
. encode ND2_trans, gen(ND2_num)

. encode B7_b_trans, gen(B7_b_num)

. encode GH1_trans, gen(GH1_num)

. encode race_trans, gen(race_num)

```

```
. encode educ_trans, gen(educ_num)
```

```
.
. svy: logistic B3_bin i.ND2_num i.B7_b_num i.GH1_num i.race_num i.educ_num
(running logistic on estimation sample)
```

Survey: Logistic regression

Number of strata =	1	Number of obs =	11,667
Number of PSUs =	11,667	Population size =	255,114,223
		Design df =	11,666
		F(17, 11650) =	57.08
		Prob > F =	0.0000

		Linearized					
	B3_bin	Odds ratio	std. err.	t	P> t	[95% conf. interval]	
ND2_num							
	2	1.085264	.1004103	0.88	0.377	.9052586	1.301062
	3	1.068285	.0911387	0.77	0.439	.9037764	1.262738
	4	1.298803	.2656084	1.28	0.201	.8698664	1.939252
	5	1.268982	.211473	1.43	0.153	.9153563	1.759224
B7_b_num							
	2	3.023388	.1478851	22.62	0.000	2.746971	3.327619
	3	6.053997	.4822353	22.61	0.000	5.178836	7.077051
	4	11.91989	4.109229	7.19	0.000	6.064588	23.42845
GH1_num							
	2	.938628	.0530243	-1.12	0.262	.8402394	1.048538
	3	1.025434	.0602294	0.43	0.669	.9139174	1.150559
	4	1.422547	.140974	3.56	0.000	1.171397	1.727545
race_num							
	2	2.031276	.1647816	8.74	0.000	1.732648	2.381373
	3	1.183977	.0847211	2.36	0.018	1.02903	1.362255
	4	1.562447	.1970896	3.54	0.000	1.220175	2.000729
	5	.9875135	.163249	-0.08	0.939	.7141897	1.36544
educ_num							



2		1.119783	.1303859	0.97	0.331	.8912739	1.40688
4		1.360402	.15108	2.77	0.006	1.094276	1.691249
S		1.178032	.1311927	1.47	0.141	.9470055	1.465419
<b>_cons</b>		.5763645	.0791819	-4.01	0.000	.4402969	.7544819

---

Note: **\_cons** estimates baseline odds.

.

All of the  $p$  – value of all categories of variable *ND2\_NUM* larger than 0.05, meaning statistically insignificant. Therefore, we have no enough evidence to conclude that long-term concerns about climate change impact current day concerns about financial stability, controlling other predictors.

(g)

```
. * (g) output data
. export delimited "K:\output.csv"
file K:\output.csv saved

.
end of do-file
```

(h)

```
library(survey)
```

Loading required package: grid

Loading required package: Matrix

Attaching package: 'Matrix'

The following objects are masked from 'package:tidyr':

expand, pack, unpack

Loading required package: survival

Attaching package: 'survey'

The following object is masked from 'package:graphics':

dotchart

```
public2022 <- read.csv("output.csv")

des <- svydesign(id = ~ CaseID, weight = ~ weight_pop, data = public2022)

fit <- svyglm(B3_bin ~ factor(ND2_num) + factor(B7_b_num) + factor(GH1_num) +
              factor(race_num) + factor(educ_num), design = des, family = quasibinomial())

summary(fit)
```

Call:

```
svyglm(formula = B3_bin ~ factor(ND2_num) + factor(B7_b_num) +
        factor(GH1_num) + factor(race_num) + factor(educ_num), design = des,
        family = quasibinomial())
```

Survey design:

```
svydesign(id = ~CaseID, weight = ~weight_pop, data = public2022)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	-0.55101	0.13738	-4.011	6.09e-05	***
factor(ND2_num)2	0.08182	0.09252	0.884	0.376515	
factor(ND2_num)3	0.06605	0.08531	0.774	0.438791	
factor(ND2_num)4	0.26144	0.20450	1.278	0.201119	
factor(ND2_num)5	0.23822	0.16665	1.429	0.152899	
factor(B7_b_num)2	1.10638	0.04891	22.619	< 2e-16	***
factor(B7_b_num)3	1.80072	0.07966	22.606	< 2e-16	***
factor(B7_b_num)4	2.47821	0.34473	7.189	6.94e-13	***
factor(GH1_num)2	-0.06334	0.05649	-1.121	0.262241	
factor(GH1_num)3	0.02512	0.05874	0.428	0.668936	
factor(GH1_num)4	0.35245	0.09910	3.557	0.000377	***
factor(race_num)2	0.70866	0.08112	8.736	< 2e-16	***
factor(race_num)3	0.16888	0.07156	2.360	0.018287	*
factor(race_num)4	0.44625	0.12614	3.538	0.000405	***

```

factor(race_num)5 -0.01257    0.16531  -0.076 0.939414
factor(educ_num)2  0.11314    0.11644   0.972 0.331254
factor(educ_num)4  0.30778    0.11106   2.771 0.005590 **
factor(educ_num)S  0.16385    0.11137   1.471 0.141255
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for quasibinomial family taken to be 1.002519)

Number of Fisher Scoring iterations: 4

pseudo_R2 <- 1 - (fit$deviance / fit$null.deviance)
pseudo_R2

[1] 0.09010785

```