

STAT 231: Problem Set 2B

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due by 5 PM on Friday, March 5

Series B homework assignments are designed to help you further ingest and practice the material covered in class over the past week(s). You are encouraged to work with other students, but all code must be written by you and you must indicate below who you discussed the assignment with (if anyone).

Steps to proceed:

1. In RStudio, go to File > Open Project, navigate to the folder with the course-content repo, select the course-content project (course-content.Rproj), and click "Open"
2. Pull the course-content repo (e.g. using the blue-ish down arrow in the Git tab in upper right window)
3. Copy ps2B.Rmd from the course repo to your repo (see page 6 of the GitHub Classroom Guide for Stat231 if needed)
4. Close the course-content repo project in RStudio
5. Open YOUR repo project in RStudio
6. In the ps2B.Rmd file in YOUR repo, replace "YOUR NAME HERE" with your name
7. Add in your responses, committing and pushing to YOUR repo in appropriate places along the way
8. Run "Knit PDF"
9. Upload the pdf to Gradescope. Don't forget to select which of your pages are associated with each problem. *You will not get credit for work on unassigned pages (e.g., if you only selected the first page but your solution spans two pages, you would lose points for any part on the second page that the grader can't see).*

If you discussed this assignment with any of your peers, please list who here:

ANSWER:

MDSR Exercise 4.14 (modified)

Use the `Pitching` data frame from the `Lahman` package to identify every pitcher in baseball history who has accumulated at least 300 wins (W) and at least 3,000 strikeouts (SO).

a. How many pitchers meet this criteria?

ANSWER: There are ten players who have met the criteria.

```
library(Lahman)
Beasts <- Pitching %>% group_by(playerID) %>% summarize(tWins = sum(W), tSO = sum(SO)) %>% filter(tWins >= 300, tSO >= 3000)
Beasts
```

```
## # A tibble: 10 x 3
##   playerID  tWins  tSO
##   <chr>    <int> <int>
## 1 carltst01    329  4136
## 2 clemereo02   354  4672
## 3 johnsra05   303  4875
## 4 johnswa01   417  3509
## 5 maddugr01   355  3371
## 6 niekrph01   318  3342
## 7 perryga01   314  3534
## 8 ryanno01    324  5714
## 9 seaveto01   311  3640
## 10 suttodo01   324  3574
```

b. Which of these pitchers had the most accumulated strikeouts? How many strikeouts had he accumulated? What is the most strikeouts he had in one season?

ANSWER: ryanno01 has the most strikeouts at 5714. He had his peak year for strikeouts in 1973, striking out batters 383 times.

```
Ryan <- Pitching %>% filter(playerID == "ryanno01") %>% arrange(desc(SO))
Ryan
```

```
##   playerID yearID stint teamID lgID  W  L  G  GS  CG  SHO  SV  IPouts   H  ER  HR
## 1 ryanno01  1973     1   CAL   AL 21 16 41 39 26   4  1   978 238 104 18
## 2 ryanno01  1974     1   CAL   AL 22 16 42 41 26   3  0   998 221 107 18
## 3 ryanno01  1977     1   CAL   AL 19 16 37 37 22   4  0   897 198  92 12
## 4 ryanno01  1972     1   CAL   AL 19 16 39 39 20   9  0   852 166  72 14
## 5 ryanno01  1976     1   CAL   AL 17 18 39 39 21   7  0   853 193 106 13
## 6 ryanno01  1989     1   TEX   AL 16 10 32 32  6   2  0   718 162  85 17
## 7 ryanno01  1987     1   HOU   NL  8 16 34 34  0   0  0   635 154  65 14
## 8 ryanno01  1978     1   CAL   AL 10 13 31 31 14   3  0   704 183  97 12
## 9 ryanno01  1982     1   HOU   NL 16 12 35 35 10   3  0   751 196  88 20
## 10 ryanno01 1990     1   TEX   AL 13  9 30 30  5   2  0   612 137  78 18
## 11 ryanno01 1988     1   HOU   NL 12 11 33 33  4   1  0   660 186  86 18
## 12 ryanno01 1979     1   CAL   AL 16 14 34 34 17   5  0   668 169  89 15
## 13 ryanno01 1985     1   HOU   NL 10 12 35 35  4   0  0   696 205  98 12
## 14 ryanno01 1991     1   TEX   AL 12  6 27 27  2   2  0   519 102  56 12
```

## 15	ryanno01	1980	1	HOU	NL	11	10	35	35	4	2	0	701	205	87	10
## 16	ryanno01	1984	1	HOU	NL	12	11	30	30	5	2	0	551	143	62	12
## 17	ryanno01	1986	1	HOU	NL	12	8	30	30	1	0	0	534	119	66	14
## 18	ryanno01	1975	1	CAL	AL	14	12	28	28	10	5	0	594	152	76	13
## 19	ryanno01	1983	1	HOU	NL	14	9	29	29	5	2	0	589	134	65	9
## 20	ryanno01	1992	1	TEX	AL	5	9	27	27	2	0	0	472	138	65	9
## 21	ryanno01	1981	1	HOU	NL	11	5	21	21	5	3	0	447	99	28	2
## 22	ryanno01	1971	1	NYN	NL	10	14	30	26	3	0	0	456	125	67	8
## 23	ryanno01	1968	1	NYN	NL	6	9	21	18	3	0	0	402	93	46	12
## 24	ryanno01	1970	1	NYN	NL	7	11	27	19	5	2	1	395	86	50	10
## 25	ryanno01	1969	1	NYN	NL	6	3	25	10	2	0	1	268	60	35	3
## 26	ryanno01	1993	1	TEX	AL	5	5	13	13	0	0	0	199	54	36	5
## 27	ryanno01	1966	1	NYN	NL	0	1	2	1	0	0	0	9	5	5	1
##	BB	SO	BAOpp	ERA	IBB	WP	HBP	BK	BFP	GF	R	SH	SF	GIDP		
## 1	162	383	0.203	2.87	2	15	7	0	1355	2	113	7	7	24		
## 2	202	367	0.190	2.89	3	9	9	0	1392	1	127	12	4	24		
## 3	204	341	0.193	2.77	7	21	9	3	1272	0	110	22	10	21		
## 4	157	329	0.171	2.28	4	18	10	0	1154	0	80	11	3	NA		
## 5	183	327	0.195	3.36	2	5	5	2	1196	0	117	13	4	12		
## 6	98	301	0.187	3.20	3	19	9	1	988	0	96	9	5	4		
## 7	87	270	0.200	2.76	2	10	4	2	873	0	75	9	1	6		
## 8	148	260	0.220	3.72	7	13	3	2	1008	0	106	11	14	18		
## 9	109	245	0.213	3.16	3	18	8	2	1050	0	100	9	3	12		
## 10	74	232	0.188	3.44	2	9	7	1	818	0	86	3	5	5		
## 11	87	228	0.227	3.52	6	10	7	7	930	0	98	10	8	7		
## 12	114	223	0.212	3.60	3	9	6	0	937	0	104	8	10	14		
## 13	95	209	0.239	3.80	8	14	9	2	983	0	108	11	12	16		
## 14	72	203	0.172	2.91	0	8	5	0	683	0	58	3	9	7		
## 15	98	200	0.236	3.35	1	10	3	1	982	0	100	7	7	17		
## 16	69	197	0.212	3.04	2	6	4	3	760	0	78	4	6	10		
## 17	82	194	0.188	3.34	5	15	4	0	729	0	72	5	4	9		
## 18	132	186	0.213	3.45	0	12	7	0	864	0	90	6	7	19		
## 19	101	183	0.195	2.98	3	5	4	1	804	0	74	7	5	20		
## 20	69	157	0.238	3.72	0	9	12	0	675	0	75	6	7	5		
## 21	68	140	0.188	1.69	1	16	1	2	605	0	34	5	3	10		
## 22	116	137	0.219	3.97	4	6	15	1	705	1	78	3	0	NA		
## 23	75	133	0.200	3.09	4	7	4	0	559	1	50	NA	NA	NA		
## 24	97	125	0.188	3.42	2	8	4	0	570	4	59	8	4	NA		
## 25	53	92	0.180	3.53	3	1	1	3	375	4	38	NA	NA	NA		
## 26	40	46	0.220	4.88	0	3	1	0	291	0	47	2	2	3		
## 27	3	6	0.350	15.00	1	1	0	0	17	0	5	NA	NA	NA		

MDSR Exercise 4.17 (modified)

- a. The Violations data set in the `mdsr` package contains information regarding the outcome of health inspections in New York City. Use these data to calculate the median violation score by zipcode and dba for zipcodes in Manhattan. What pattern (if any) do you see between the number of inspections and the median score? Generate a visualization to support your response.

ANSWER: There isn't a very noticeable association. Although you can see that as inspection count increases past a certain point, there are no very low scores since these people most likely have to maintain as they are held more liable. This is why variation shrinks dramatically in the range of scores as inspection count increases.

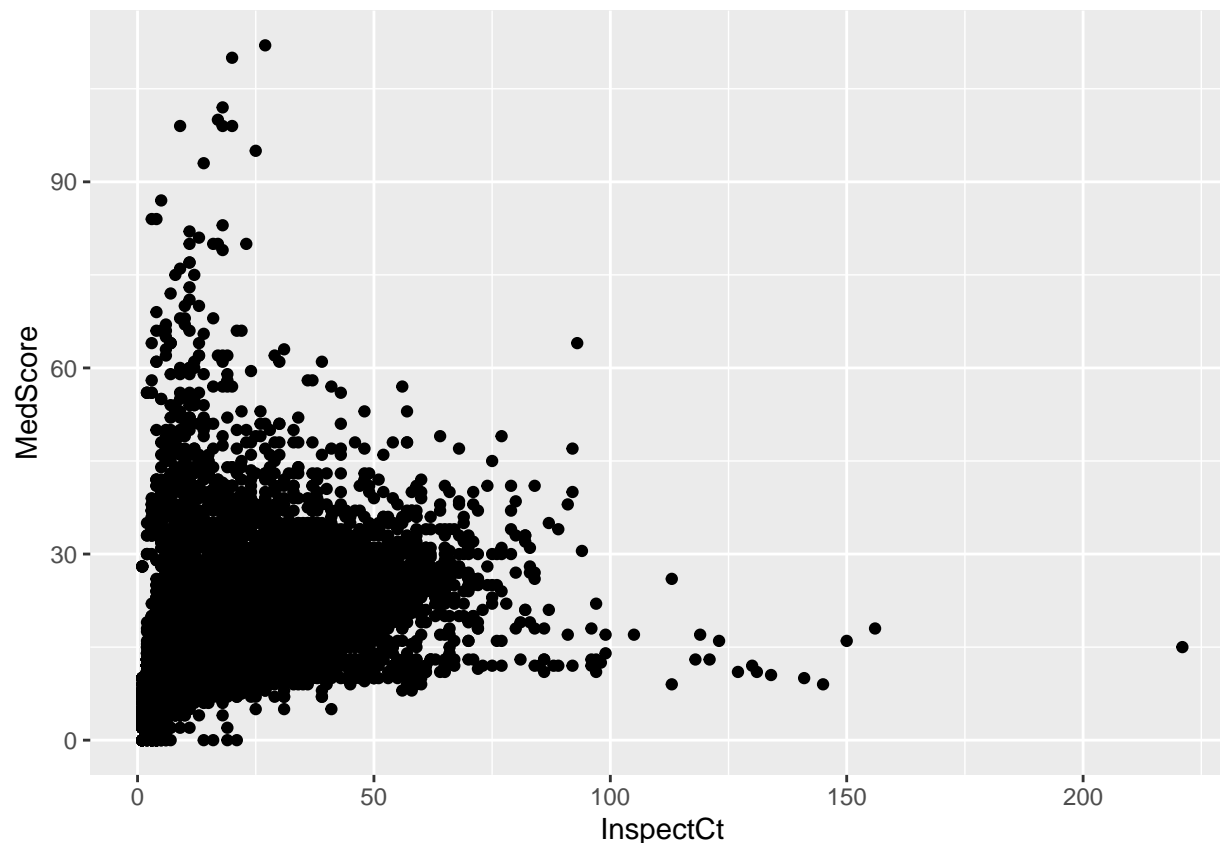
```
library(mdsr)
Violate <- Violations %>% filter(!is.na(score)) %>% group_by(dba, zipcode) %>% summarize(InspectCt = n())
```

'summarise()' has grouped output by 'dba'. You can override using the '.groups' argument.

```
inspectionmap <- ggplot(data = Violate) + geom_point(mapping = aes(y = MedScore, x = InspectCt))
Violate
```

```
## # A tibble: 23,353 x 4
## # Groups:   dba [19,758]
##   dba                zipcode InspectCt MedScore
##   <chr>              <int>    <int>    <dbl>
## 1 ''W'' CAFE        10018         23         22
## 2 (LEWIS DRUG STORE) LOCANDA VINI E OLII  11238         17         20
## 3 (LIBRARY) FOUR & TWENTY BLACKBIRDS     11238          9          9
## 4 (PUBLIC FARE) 81st street and central park west (~ 10019         19         19
## 5 @NINE             10036         50         14
## 6 / L'ECOLE         10013         15         19
## 7 #1 GARDEN CHINESE 11215         24         21
## 8 #1 SABOR LATINO RESTAURANT 10466         39         21
## 9 $1 PIZZA $2 BEER  10012         40         17
## 10 1 2 3 BURGER SHOT BEER 10019         18         20
## # ... with 23,343 more rows
```

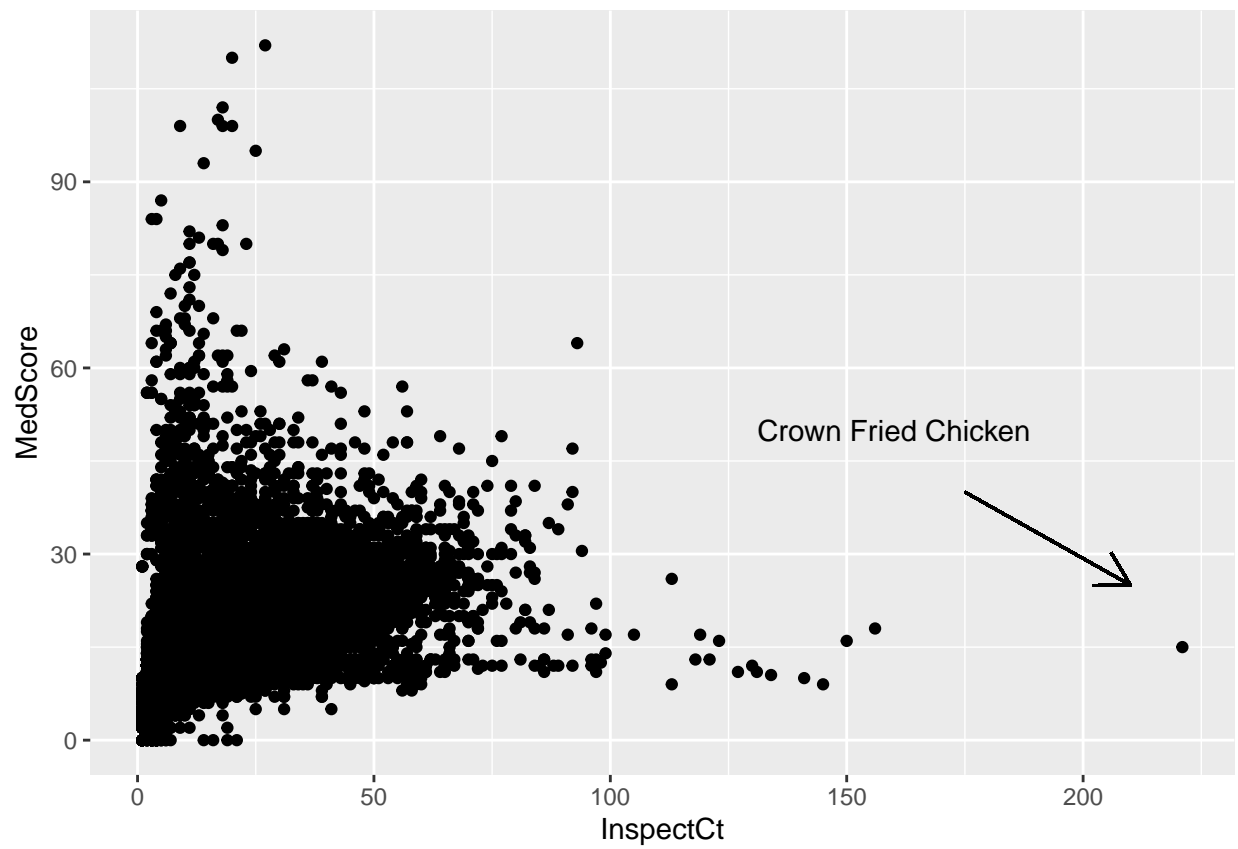
```
inspectionmap
```



- b. In your visualization in part (a), there should be at least a few points that stand out as outliers. For *one of the outliers*, add text to the outlier identifying what business it is and an arrow pointing from the text to the observation. First, you may want to **filter** to identify the name of the business (so you know what text to add to the plot).

(Can't remember how to create a curved arrow in `ggplot`? The answers to this question on Stack Exchange may help. Can't remember how to add text to the plot in `ggplot`? Check out the text examples with `annotate` here, or answers to this question that use `geom_text`.)

```
inspectionmap+ geom_segment(aes(x = 175, y = 40, xend = 210, yend = 25),
                             arrow = arrow(length = unit(0.5, "cm")))) + annotate("text", x = 160, y = 50, label =
```



MDSR Exercise 5.7

Generate the code to convert the data frame shown with this problem in the textbook (on page 130, and shown below) to wide format (i.e., the result table). Hint: use `gather()` in conjunction with `spread()`; OR `pivot_longer()` in conjunction with `pivot_wider()`.

```
FakeDataLong <- data.frame(grp = c("A","A","B", "B")
  , sex = c("F", "M", "F", "M")
  , meanL = c(0.22, 0.47, 0.33, 0.55)
  , sdL = c(0.11, 0.33, 0.11, 0.31)
  , meanR = c(0.34, 0.57, 0.40, 0.65)
  , sdR = c(0.08, 0.33, 0.07, 0.27))
```

FakeDataLong

```
##   grp sex meanL  sdL meanR  sdR
## 1  A  F  0.22 0.11  0.34 0.08
## 2  A  M  0.47 0.33  0.57 0.33
## 3  B  F  0.33 0.11  0.40 0.07
## 4  B  M  0.55 0.31  0.65 0.27
```

```
library(stringr)
FakeDataShort <- FakeDataLong %>%
  mutate(NmeanL = str_c(sex, "meanL"), NsdL = str_c(sex, "sdL"), NmeanR = str_c(sex, "meanR"), NsdR = str_c(sex, "sdR"))
FakeDataShort[2] <- NULL
FakeDataShort <- FakeDataShort %>%
  pivot_wider(names_from = NmeanL, values_from = meanL, names_sep="") %>%
  pivot_wider(names_from = NsdL, values_from = sdL, names_sep="") %>%
  pivot_wider(names_from = NmeanR, values_from = meanR, names_sep="") %>%
  pivot_wider(names_from = NsdR, values_from = sdR, names_sep="")

FakeDataShort %>% group_by(grp) %>% summarize(F.meanL=max(FmeanL,na.rm=TRUE), M.meanL=max(MmeanL,na.rm=TRUE), F.meanR=max(FmeanR,na.rm=TRUE), M.meanR=max(MmeanR,na.rm=TRUE), F.sdL=max(FsdL,na.rm=TRUE), M.sdL=max(MsdL,na.rm=TRUE), F.sdR=max(FsdR,na.rm=TRUE), M.sdR=max(MsdR,na.rm=TRUE))
```

```
## # A tibble: 2 x 9
##   grp  F.meanL M.meanL F.sdL M.sdL F.meanR M.meanR F.sdR M.sdR
## * <chr>   <dbl>   <dbl> <dbl> <dbl>   <dbl>   <dbl> <dbl> <dbl>
## 1 A      0.22    0.47  0.11  0.33    0.34    0.570  0.08  0.33
## 2 B      0.33    0.55  0.11  0.31    0.4     0.65  0.07  0.27
```


PUG Brainstorming

What topics or questions are you interested in exploring related to your PUG theme? Dream big here. Don't worry about whether there is data out there that's available and accessible that you could use to address your questions/topics. Just brainstorm some ideas that get you excited. Then, email your PUG team with your ideas. Title the email "PS2B Brainstorming: PUG [#] [Topic]" and CC me (kcorreia@amherst.edu) on the email. If another PUG member already initiated the email, reply all to their email.

If you don't remember your PUG # and Topic, please see the file "PUGs" on the Moodle page under this week.

If you don't know your PUG members email address, go to the class's Google group conversations (e.g., by clicking the link "Link to Google group conversations" at the top of our Moodle course page). Then, on the navigation panel (left hand side), select "Members".

ANSWER: Do not write anything here. Email your ideas to your PUG team and me in a message titled "PS2B Brainstorming: PUG [#] [Topic]".