

class17

Moises Gonzalez A17579866

Pertussis, or whooping cough, is a highly contagious lung infection caused by a bacteria *B. pertussis*.

The CDC tracks reported cases in the U.S. since the 1920s.

```
cdc <- data.frame(data.frame(
  Year = c(1922L, 1923L,
           1924L, 1925L,
           1926L, 1927L, 1928L,
           1929L, 1930L, 1931L,
           1932L, 1933L, 1934L,
           1935L, 1936L, 1937L,
           1938L, 1939L, 1940L,
           1941L, 1942L, 1943L,
           1944L, 1945L,
           1946L, 1947L, 1948L,
           1949L, 1950L, 1951L,
           1952L, 1953L, 1954L,
           1955L, 1956L, 1957L,
           1958L, 1959L, 1960L,
           1961L, 1962L, 1963L,
           1964L, 1965L,
           1966L, 1967L, 1968L,
           1969L, 1970L, 1971L,
           1972L, 1973L, 1974L,
           1975L, 1976L, 1977L,
           1978L, 1979L, 1980L,
           1981L, 1982L, 1983L,
           1984L, 1985L,
           1986L, 1987L, 1988L,
           1989L, 1990L, 1991L,
           1992L, 1993L, 1994L,
```

```

1995L, 1996L, 1997L,
1998L, 1999L, 2000L,
2001L, 2002L, 2003L,
2004L, 2005L,
2006L, 2007L, 2008L,
2009L, 2010L, 2011L,
2012L, 2013L, 2014L,
2015L, 2016L, 2017L,
2018L, 2019L, 2020L,
2021L),
No..Reported.Pertussis.Cases = c(107473,
164191, 165418, 152003,
202210, 181411,
161799, 197371, 166914,
172559, 215343,
179135, 265269, 180518,
147237, 214652,
227319, 103188, 183866,
222202, 191383,
191890, 109873, 133792,
109860, 156517,
74715, 69479, 120718,
68687, 45030, 37129,
60886, 62786, 31732,
28295, 32148, 40005,
14809, 11468, 17749,
17135, 13005, 6799,
7717, 9718, 4810,
3285, 4249, 3036,
3287, 1759, 2402, 1738,
1010, 2177, 2063,
1623, 1730, 1248,
1895, 2463, 2276, 3589,
4195, 2823, 3450,
4157, 4570, 2719,
4083, 6586, 4617, 5137,
7796, 6564, 7405,
7298, 7867, 7580,
9771, 11647, 25827,
25616, 15632, 10454,
13278, 16858, 27550,

```

```
18719,48277,28639,  
32971,20762,17972,  
18975,15609,  
18617,6124,2116)
```

```
))
```

We can now plot the number of reported pertussis cases per year in the U.S.

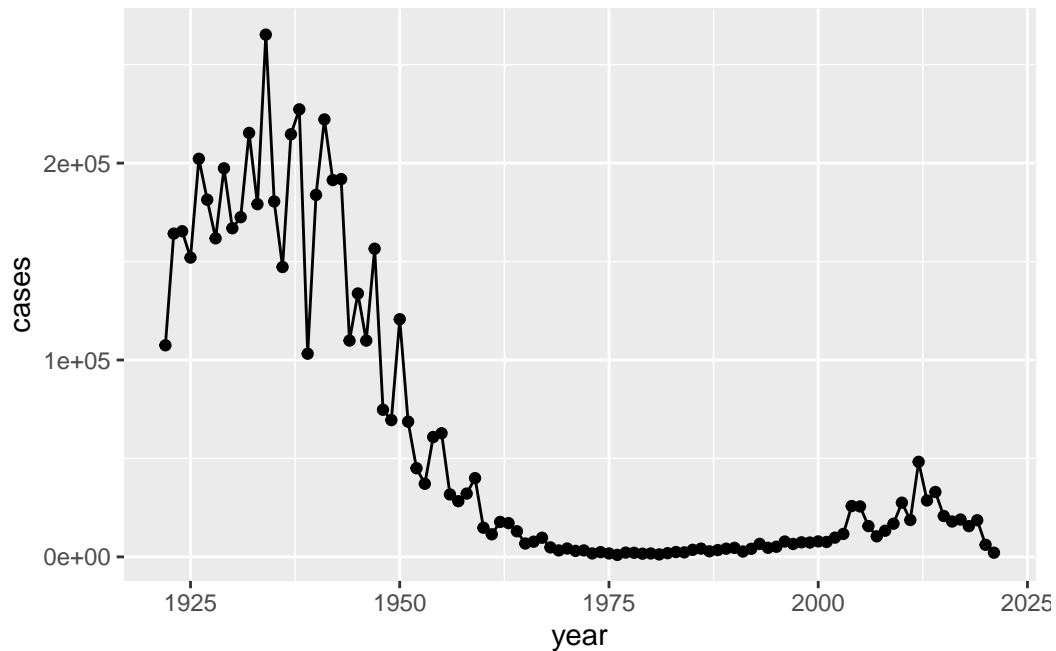
```
library(ggplot2)  
  
ggplot(cdc) +  
  aes(cdc$Year, cdc$No..Reported.Pertussis.Cases)+  
  geom_point()+  
  geom_line()+  
  labs(x="year", y="cases")
```

```
Warning: Use of `cdc$Year` is discouraged.  
i Use `Year` instead.
```

```
Warning: Use of `cdc$No..Reported.Pertussis.Cases` is discouraged.  
i Use `No..Reported.Pertussis.Cases` instead.
```

```
Warning: Use of `cdc$Year` is discouraged.  
i Use `Year` instead.
```

```
Warning: Use of `cdc$No..Reported.Pertussis.Cases` is discouraged.  
i Use `No..Reported.Pertussis.Cases` instead.
```



The first big “Whole-cell” pertussis vaccine was introduced in 1942

```
colnames(cdc) <- c('Year', 'Cases')
```

```
head(cdc)
```

	Year	Cases
1	1922	107473
2	1923	164191
3	1924	165418
4	1925	152003
5	1926	202210
6	1927	181411

```
ggplot(cdc) +
  aes(cdc$Year, cdc$Cases)+
  geom_point()+
  geom_line()+
  labs(x="Year", y="Cases")+
  geom_vline(xintercept=1942, col="blue")+
  geom_vline(xintercept=1980, col="grey", linetype=2)+
```

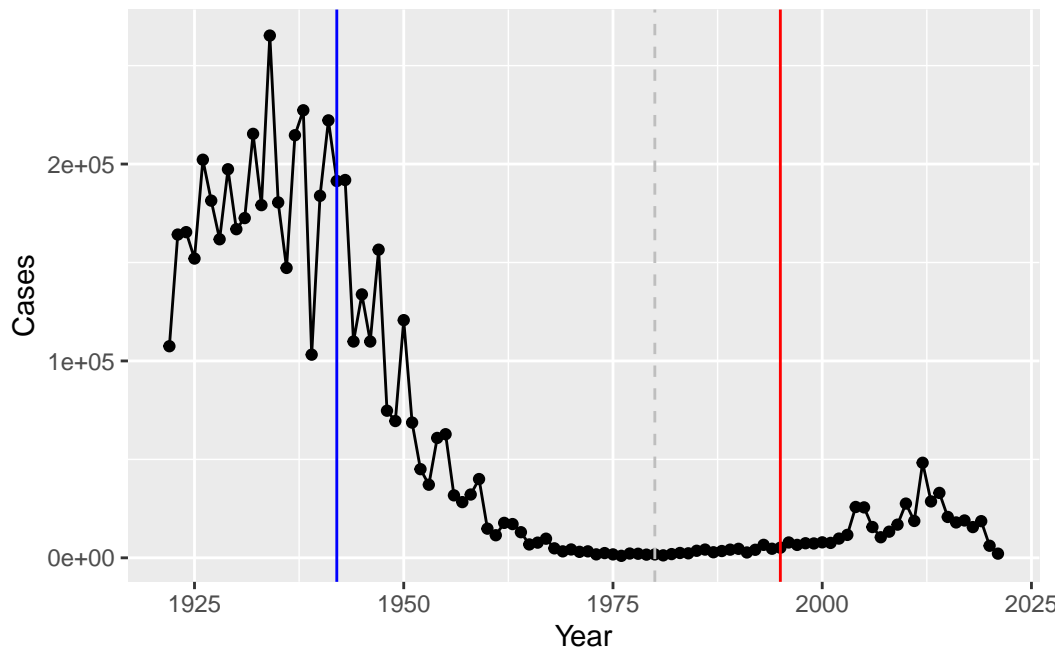
```
geom_vline(xintercept=1995, col="red")
```

Warning: Use of `cdc\$Year` is discouraged.
i Use `Year` instead.

Warning: Use of `cdc\$Cases` is discouraged.
i Use `Cases` instead.

Warning: Use of `cdc\$Year` is discouraged.
i Use `Year` instead.

Warning: Use of `cdc\$Cases` is discouraged.
i Use `Cases` instead.



Something bi is happening with pertussis cases and big outbreaks are once again a major public health concern! BUGGER

One of the main hypothesis for the increasing case numbers is waning vaccine efficacy with the newer vaccine.

Enter the CMI-PB project, which is studying this problem on a large scale. Let's see what data they have.

Their data is available in JSON format ("key: value" pair style). We will use the "jsonlite" package to read their data

```
library(jsonlite)

subject <- read_json("https://www.cmi-pb.org/api/subject", simplifyVector = TRUE)

head(subject)
```

```
  subject_id infancy_vac biological_sex ethnicity race
1          1          wP    Female Not Hispanic or Latino White
2          2          wP    Female Not Hispanic or Latino White
3          3          wP    Female           Unknown White
4          4          wP    Male Not Hispanic or Latino Asian
5          5          wP    Male Not Hispanic or Latino Asian
6          6          wP    Female Not Hispanic or Latino White
 year_of_birth date_of_boost   dataset
1 1986-01-01    2016-09-12 2020_dataset
2 1968-01-01    2019-01-28 2020_dataset
3 1983-01-01    2016-10-10 2020_dataset
4 1988-01-01    2016-08-29 2020_dataset
5 1991-01-01    2016-08-29 2020_dataset
6 1988-01-01    2016-10-10 2020_dataset
```

Q4. How many aP and wP infancy vaccinated subjects are in the dataset?

```
table(subject$infancy_vac)
```

```
aP wP
47 49
```

Q5. How many Male and Female subjects/patients are in the dataset?

```
table(subject$biological_sex)
```

```
Female  Male
   66    30
```

Q6. What is the breakdown of race and biological sex (e.g. number of Asian females, White males etc...)?

```
table(subject$race)
```

```

American Indian/Alaska Native
                        1
Asian
                        27
Black or African American
                        2
More Than One Race
                        10
Native Hawaiian or Other Pacific Islander
                        2
Unknown or Not Reported
                        14
White
                        40
```

```
table(subject$race, subject$biological_sex)
```

	Female	Male
American Indian/Alaska Native	0	1
Asian	18	9
Black or African American	2	0
More Than One Race	8	2
Native Hawaiian or Other Pacific Islander	1	1
Unknown or Not Reported	10	4
White	27	13

Now let's read some more data tables from CMI-PB:

```
specimen <- read_json("http://cmi-pb.org/api/specimen", simplifyVector = TRUE)
head(specimen)
```

```
specimen_id subject_id actual_day_relative_to_boost
```

1	1	1	-3
2	2	1	736
3	3	1	1
4	4	1	3
5	5	1	7
6	6	1	11

	planned_day_relative_to_boost	specimen_type	visit
1	0	Blood	1
2	736	Blood	10
3	1	Blood	2
4	3	Blood	3
5	7	Blood	4
6	14	Blood	5

I want to “join” (a.k.a. “merge”/link/etc.) the `subject` and `specimen` tables together. I will use the **dplyr** package for this.

```
library(dplyr)
```

Attaching package: 'dplyr'

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

`filter`, `lag`

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

`intersect`, `setdiff`, `setequal`, `union`

```
meta <- inner_join(subject, specimen)
```

Joining with `by = join_by(subject_id)`

```
head(meta)
```


	subject_id	infancy_vac	biological_sex	ethnicity	race
1	1	wP	Female Not Hispanic or Latino	White	
2	1	wP	Female Not Hispanic or Latino	White	
3	1	wP	Female Not Hispanic or Latino	White	
4	1	wP	Female Not Hispanic or Latino	White	
5	1	wP	Female Not Hispanic or Latino	White	
6	1	wP	Female Not Hispanic or Latino	White	

	year_of_birth	date_of_boost	dataset	specimen_id
1	1986-01-01	2016-09-12	2020_dataset	1
2	1986-01-01	2016-09-12	2020_dataset	2
3	1986-01-01	2016-09-12	2020_dataset	3
4	1986-01-01	2016-09-12	2020_dataset	4
5	1986-01-01	2016-09-12	2020_dataset	5
6	1986-01-01	2016-09-12	2020_dataset	6

	actual_day_relative_to_boost	planned_day_relative_to_boost	specimen_type
1	-3	0	Blood
2	736	736	Blood
3	1	1	Blood
4	3	3	Blood
5	7	7	Blood
6	11	14	Blood

	visit
1	1
2	10
3	2
4	3
5	4
6	5

```
colnames(subject)
```

```
[1] "subject_id"      "infancy_vac"     "biological_sex"  "ethnicity"
[5] "race"           "year_of_birth"   "date_of_boost"   "dataset"
```

```
colnames(specimen)
```

```
[1] "specimen_id"      "subject_id"
[3] "actual_day_relative_to_boost" "planned_day_relative_to_boost"
[5] "specimen_type"     "visit"
```

```
colnames(meta)
```

```
[1] "subject_id"           "infancy_vac"
[3] "biological_sex"       "ethnicity"
[5] "race"                 "year_of_birth"
[7] "date_of_boost"        "dataset"
[9] "specimen_id"          "actual_day_relative_to_boost"
[11] "planned_day_relative_to_boost" "specimen_type"
[13] "visit"
```

```
ab <- read_json("http://cmi-pb.org/api/ab_titer",
                 simplifyVector = TRUE)
head(ab)
```

	specimen_id	isotype	is_antigen_specific	antigen	MFI	MFI_normalised
1	1	IgE	FALSE	Total	1110.21154	2.493425
2	1	IgE	FALSE	Total	2708.91616	2.493425
3	1	IgG	TRUE	PT	68.56614	3.736992
4	1	IgG	TRUE	PRN	332.12718	2.602350
5	1	IgG	TRUE	FHA	1887.12263	34.050956
6	1	IgE	TRUE	ACT	0.10000	1.000000

	unit	lower_limit_of_detection
1	UG/ML	2.096133
2	IU/ML	29.170000
3	IU/ML	0.530000
4	IU/ML	6.205949
5	IU/ML	4.679535
6	IU/ML	2.816431

Now I can join meta that we made above and contained all info about the subject and specimens with this ab data.

```
abdata <- inner_join(meta, ab)
```

Joining with `by = join_by(specimen_id)`

```
dim(abdata)
```

[1] 32675 20

Q11. How many specimens (i.e. entries in abdata) do we have for each isotype?

```
table(abdata$isotype)
```

```
 IgE  IgG IgG1 IgG2 IgG3 IgG4
6698 1413 6141 6141 6141 6141
```

Q12. What do you notice about the number of visit 8 specimens compared to other visits?

```
table(abdata$visit)
```

```
 1    2    3    4    5    6    7    8
5795 4640 4640 4640 4640 4320 3920  80
```

There are way less visit 8 specimens because the project is still ongoing and we have not got that data for all individuals yet.

Examine IgG1 Ab titer levels

We will use the `filter()` function from `dplyr` to focus on just IgG1 isotope and visits 1 to 7 (i.e. exclude visit 8 as there are not as many specimen there yet).

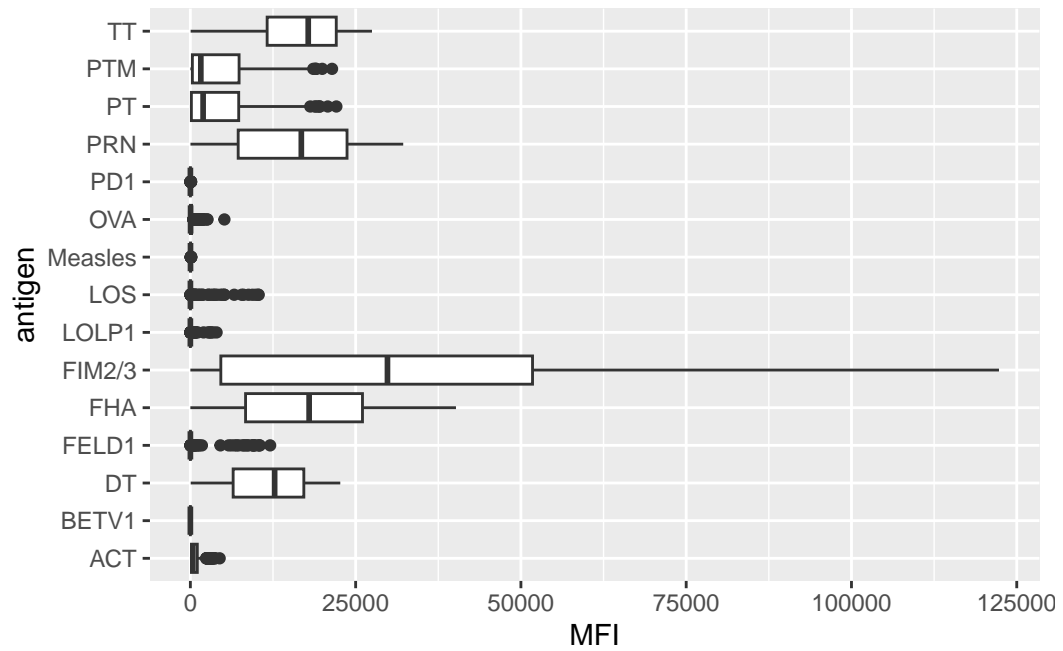
```
ig1 <- filter(abdata, isotype == "IgG1", visit!=8)
head(ig1)
```

```
subject_id infancy_vac biological_sex ethnicity race
1          1          wP      Female Not Hispanic or Latino White
2          1          wP      Female Not Hispanic or Latino White
3          1          wP      Female Not Hispanic or Latino White
4          1          wP      Female Not Hispanic or Latino White
5          1          wP      Female Not Hispanic or Latino White
6          1          wP      Female Not Hispanic or Latino White
 year_of_birth date_of_boost   dataset specimen_id
1  1986-01-01   2016-09-12 2020_dataset          1
2  1986-01-01   2016-09-12 2020_dataset          1
```

3	1986-01-01	2016-09-12	2020_dataset	1			
4	1986-01-01	2016-09-12	2020_dataset	1			
5	1986-01-01	2016-09-12	2020_dataset	1			
6	1986-01-01	2016-09-12	2020_dataset	1			
	actual_day_relative_to_boost	planned_day_relative_to_boost	specimen_type				
1		-3	0	Blood			
2		-3	0	Blood			
3		-3	0	Blood			
4		-3	0	Blood			
5		-3	0	Blood			
6		-3	0	Blood			
	visit	isotype	is_antigen_specific	antigen	MFI	MFI_normalised	unit
1	1	IgG1	TRUE	ACT	274.355068	0.6928058	IU/ML
2	1	IgG1	TRUE	LOS	10.974026	2.1645083	IU/ML
3	1	IgG1	TRUE	FELD1	1.448796	0.8080941	IU/ML
4	1	IgG1	TRUE	BETV1	0.100000	1.0000000	IU/ML
5	1	IgG1	TRUE	LOLP1	0.100000	1.0000000	IU/ML
6	1	IgG1	TRUE	Measles	36.277417	1.6638332	IU/ML
	lower_limit_of_detection						
1					3.848750		
2					4.357917		
3					2.699944		
4					1.734784		
5					2.550606		
6					4.438966		

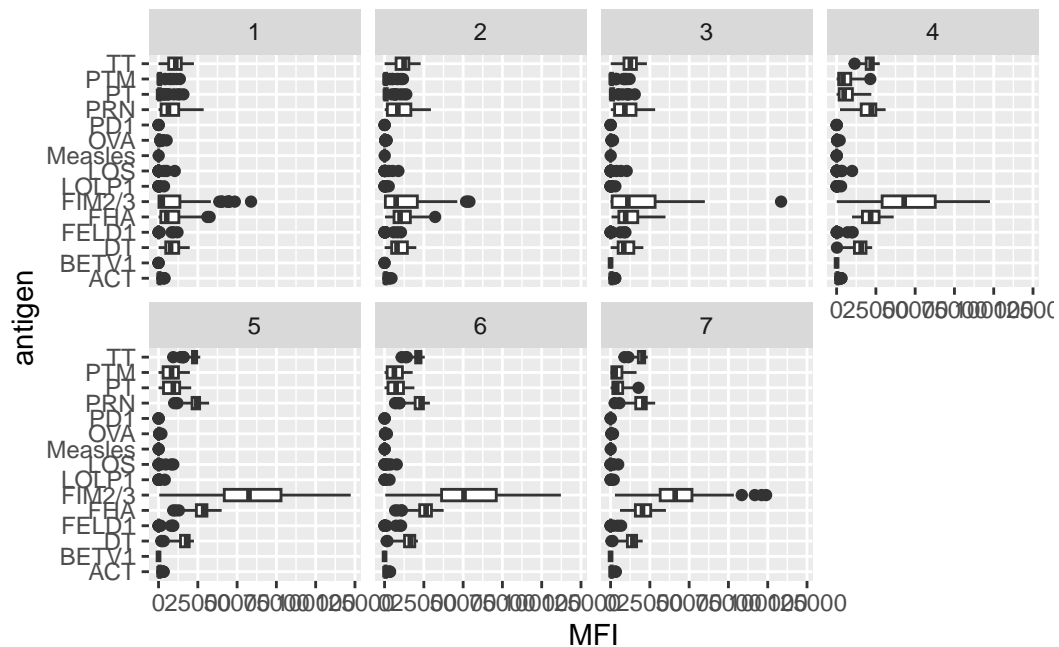
Q13. Complete the following code to make a summary boxplot of Ab titer levels for all antigens:

```
ggplot(ig1) +
  aes(MFI, antigen) +
  geom_boxplot()
```



and facet by visit

```
ggplot(ig1) +
  aes(MFI, antigen) +
  geom_boxplot() +
  facet_wrap(vars(visit), nrow=2)
```

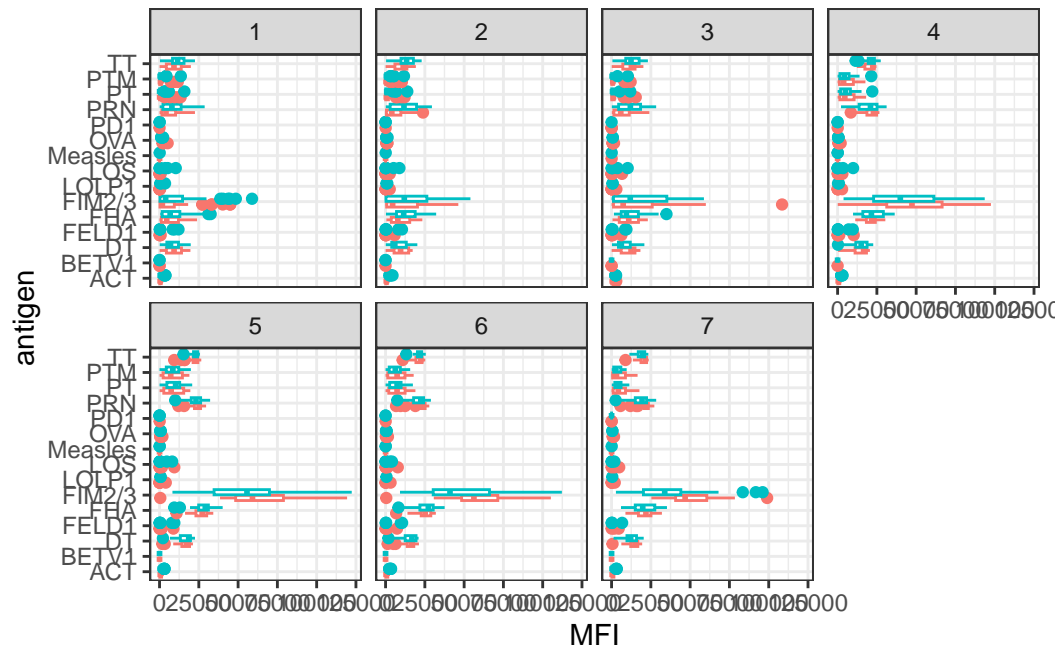


Clearly FIM2/3 changes. This is “Fimbrial protein” that makes the bacteria pilus and is involved in cell adhesion.

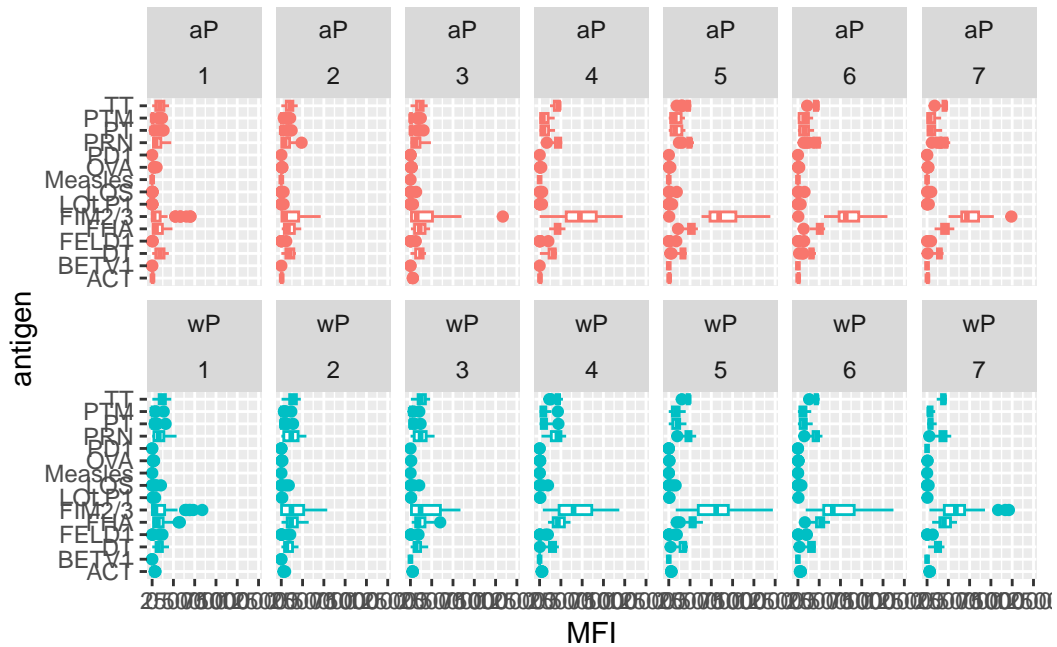
PT Pertussis toxin

FHA is Filamentous hemagglutain

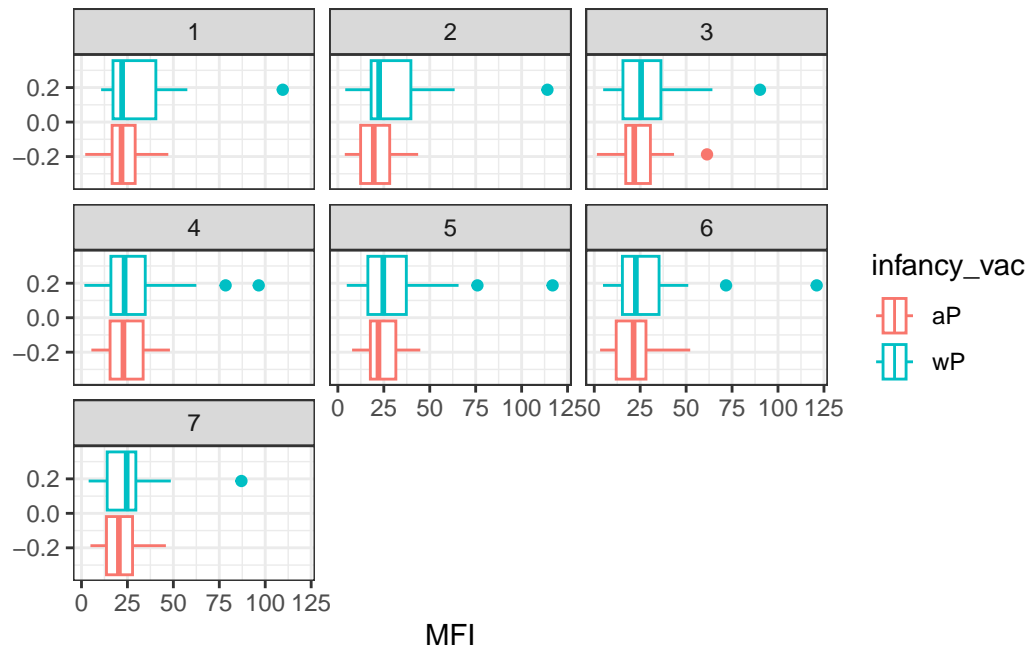
```
ggplot(ig1) +
  aes(MFI, antigen, col=infancy_vac ) +
  geom_boxplot(show.legend = FALSE) +
  facet_wrap(vars(visit), nrow=2) +
  theme_bw()
```



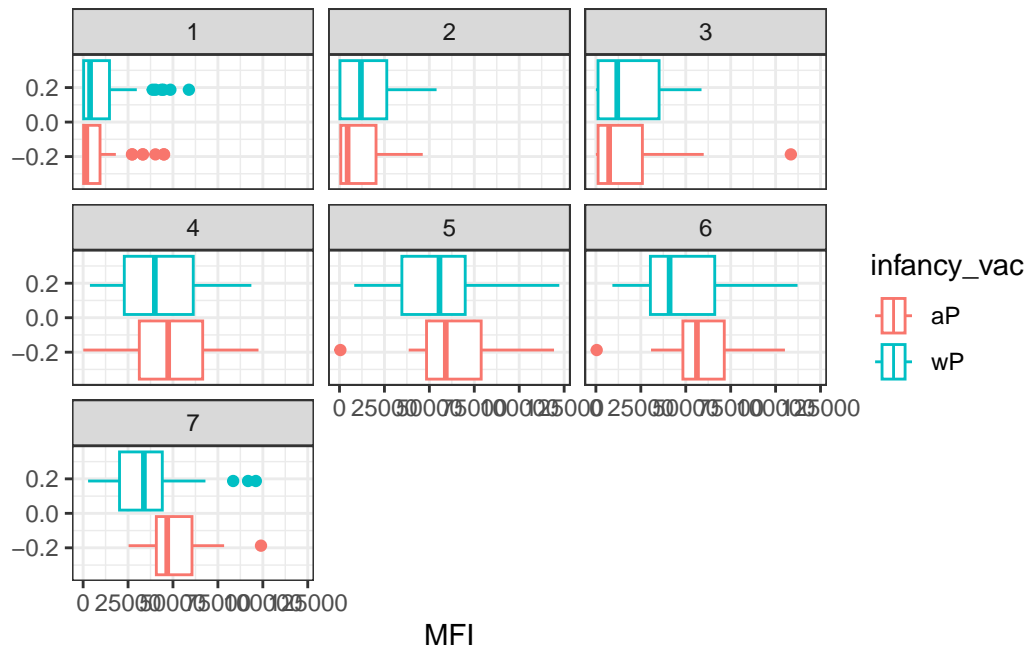
```
ggplot(ig1) +
  aes(MFI, antigen, col=infancy_vac ) +
  geom_boxplot(show.legend = FALSE) +
  facet_wrap(vars(infancy_vac, visit), nrow=2)
```



```
filter(ig1, antigen=="Measles") %>%
  ggplot() +
  aes(MFI, col=infancy_vac) +
  geom_boxplot(show.legend = TRUE) +
  facet_wrap(vars(visit)) +
  theme_bw()
```

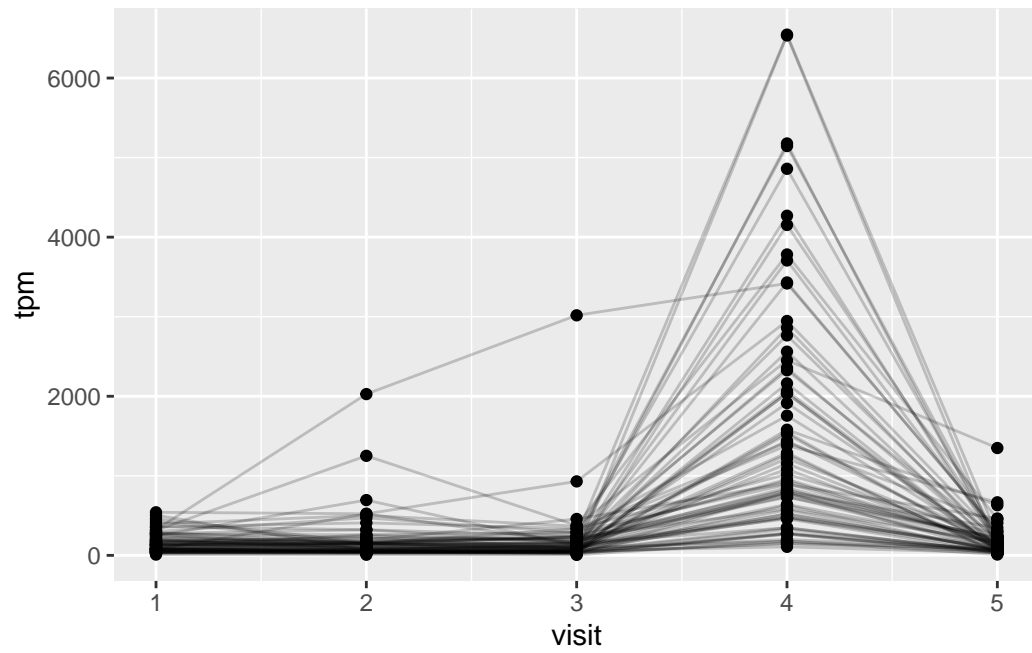
```
filter(ig1, antigen=="FIM2/3") %>%
  ggplot() +
  aes(MFI, col=infancy_vac) +
  geom_boxplot(show.legend = TRUE) +
  facet_wrap(vars(visit)) +
  theme_bw()
```



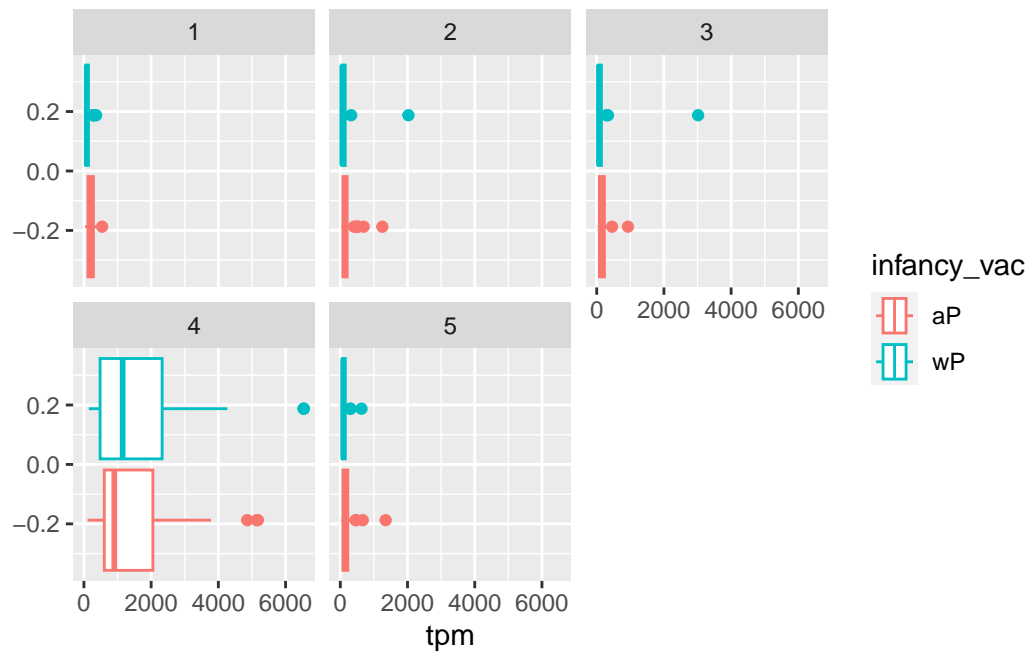
```
url <- "https://www.cmi-pb.org/api/v2/rnaseq?versioned_ensembl_gene_id=eq.ENSOG00000211896."
rna <- read_json(url, simplifyVector = TRUE)
#meta <- inner_join(specimen, subject)
ssrna <- inner_join(rna, meta)
```

Joining with `by = join_by(specimen_id)`

```
rna <- read_json(url, simplifyVector = TRUE)
ggplot(ssrna) +
  aes(visit, tpm, group=subject_id) +
  geom_point() +
  geom_line(alpha=0.2)
```



```
ggplot(ssrna) +
  aes(tpm, col=infancy_vac) +
  geom_boxplot() +
  facet_wrap(vars(visit))
```



```
ssrna %>%
  filter(visit==4) %>%
  ggplot() +
    aes(tpm, col=infancy_vac) + geom_density() +
    geom_rug()
```

