

Class17

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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.

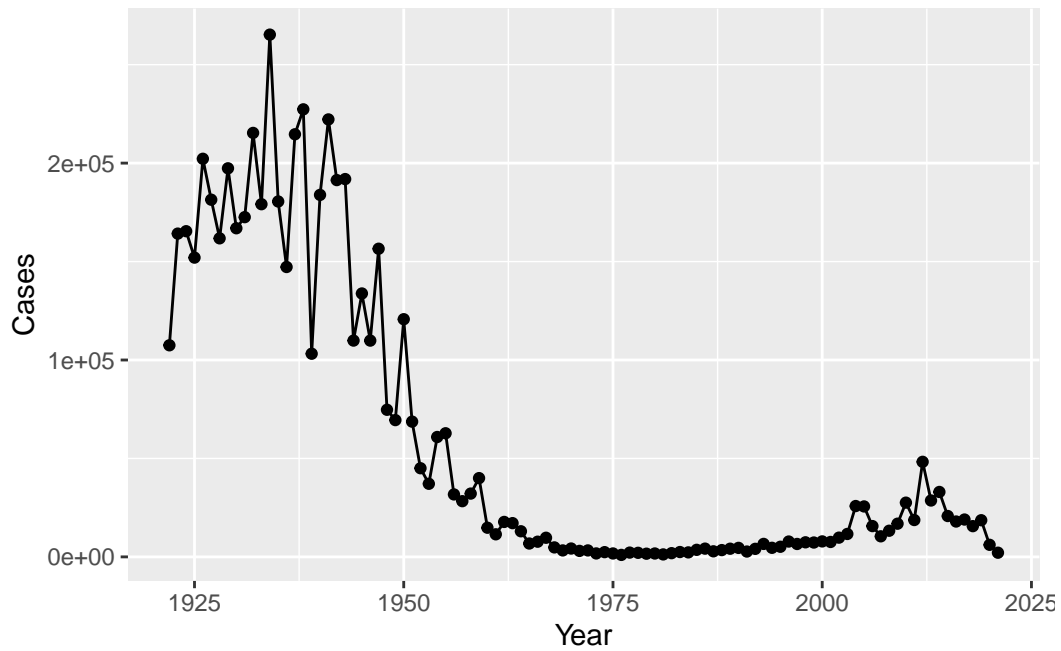
```
cdc <- data.frame(
  Year = c(1922L, 1923L, 1924L, 1925L,
           1926L, 1927L, 1928L, 1929L, 1930L, 1931L,
           1932L, 1933L, 1934L, 1935L, 1936L,
           1937L, 1938L, 1939L, 1940L, 1941L, 1942L,
           1943L, 1944L, 1945L, 1946L, 1947L,
           1948L, 1949L, 1950L, 1951L, 1952L,
           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),
  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)
```

```
)
```

Q1.

```
library(ggplot2)  
  
ggplot(cdc) +  
  aes(Year, Cases) +  
  geom_point() +  
  geom_line() +  
  labs(x="Year", y="Cases")
```

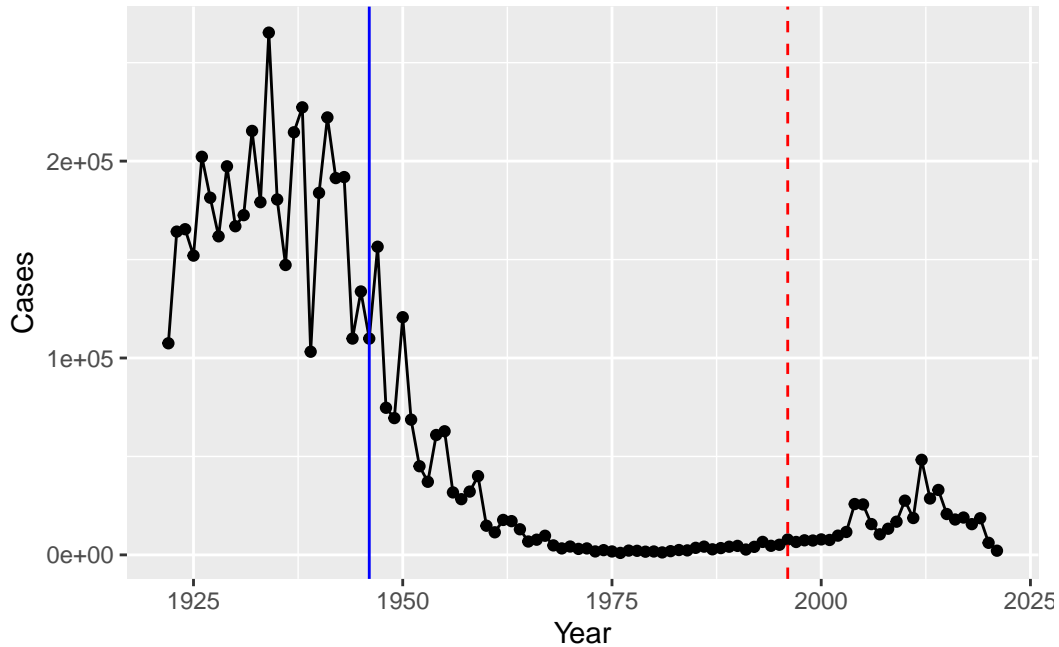


The first big “whole-cell” pertussis vaccine program started in 1942

Q2.

```
library(ggplot2)

ggplot(cdc) +
  aes(Year, Cases) +
  geom_point() +
  geom_line() +
  geom_vline(xintercept=1946, col="blue") +
  geom_vline(xintercept=1996, col="red", linetype=2)
```



Q3. Describe what happened after the introduction of the aP vaccine? Do you have a possible explanation for the observed trend?

We could possibly have a better system of detection of the disease, parents may not vaccinate children because they think that vaccinations for TDaP and other vaccines bring neurological diseases, or the bacterium became stronger and the vaccine is no longer effective to destroy it.

Something 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 aP vaccine.

Enter the CMI-PB project, which is studying this problem on 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, 3)
```

	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

	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

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?

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

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

```
Female   Male
    66    30
```

```
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

Let's read more database tables from CMI-PB:

```
specimen <- read_json("https://www.cmi-pb.org/api/specimen",
                      simplifyVector = TRUE)
head(specimen, 3)
```

	specimen_id	subject_id	actual_day_relative_to_boost	
1	1	1	-3	
2	2	1	736	
3	3	1	1	

	planned_day_relative_to_boost	specimen_type	visit
1	0	Blood	1
2	736	Blood	10
3	1	Blood	2

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

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

Now I can join the “meta” that we made above and contains all info about the subjects and specimens with this ab data.

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

Joining with `by = join_by(specimen_id)`

```
head(abdata)
```

	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	IgE	FALSE	Total	1110.21154	2.493425	UG/ML
2	1	IgE	FALSE	Total	2708.91616	2.493425	IU/ML

3	1	IgG	TRUE	PT	68.56614	3.736992	IU/ML
4	1	IgG	TRUE	PRN	332.12718	2.602350	IU/ML
5	1	IgG	TRUE	FHA	1887.12263	34.050956	IU/ML
6	1	IgE	TRUE	ACT	0.10000	1.000000	IU/ML

	lower_limit_of_detection
1	2.096133
2	29.170000
3	0.530000
4	6.205949
5	4.679535
6	2.816431

```
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 isotype and visits 1 to 7 (i.e. examine our)

```
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	

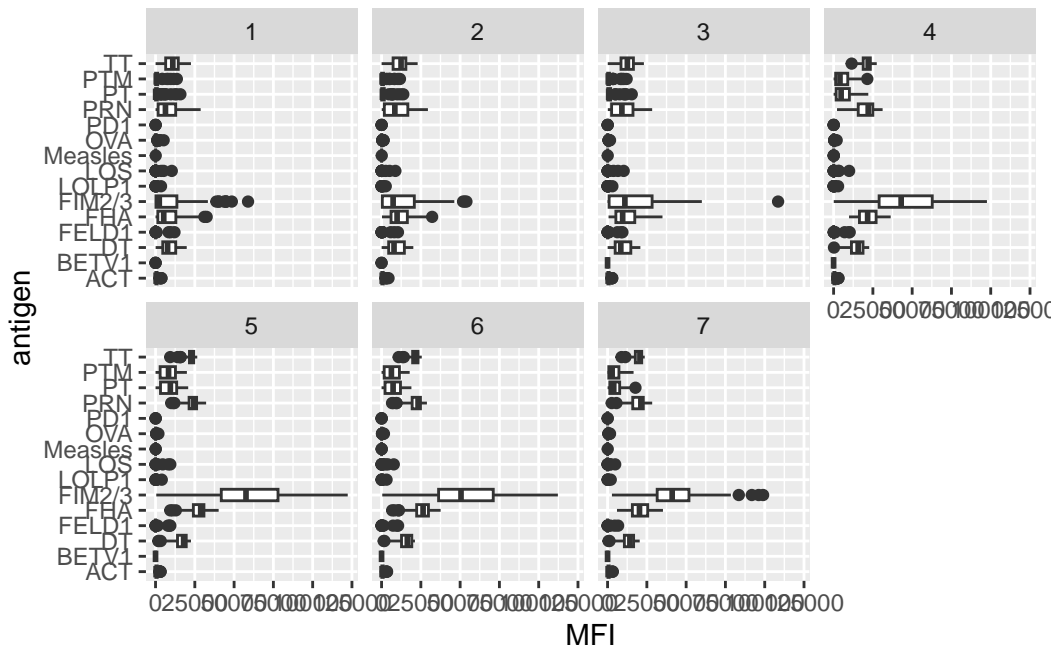
	year_of_birth	date_of_boost	dataset	specimen_id
1	1986-01-01	2016-09-12	2020_dataset	1
2	1986-01-01	2016-09-12	2020_dataset	1
3	1986-01-01	2016-09-12	2020_dataset	1
4	1986-01-01	2016-09-12	2020_dataset	1
5	1986-01-01	2016-09-12	2020_dataset	1
6	1986-01-01	2016-09-12	2020_dataset	1

	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

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

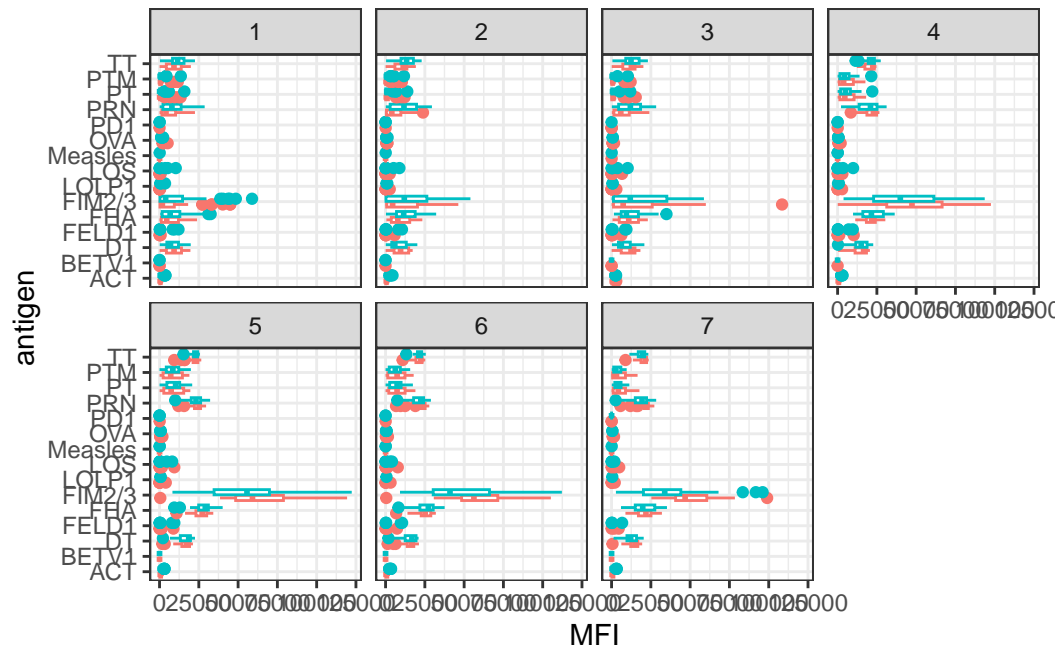


Q14. What antigens show differences in the level of IgG1 antibody titers recognizing them over time? Why these and not others?

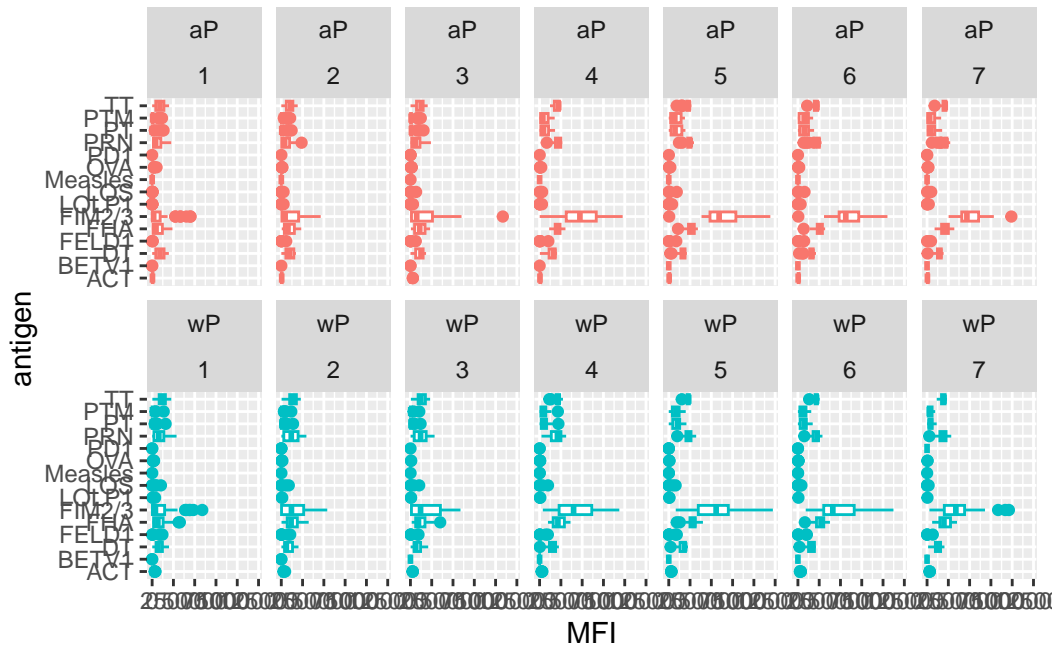
FIM2/3 show differences in MFI. This is “Fimbrial protein” that makes the bacteria pilus and is involved in cell adhesion.

PT Pertussis Toxin FHA Filamentous Hemagglutinin

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

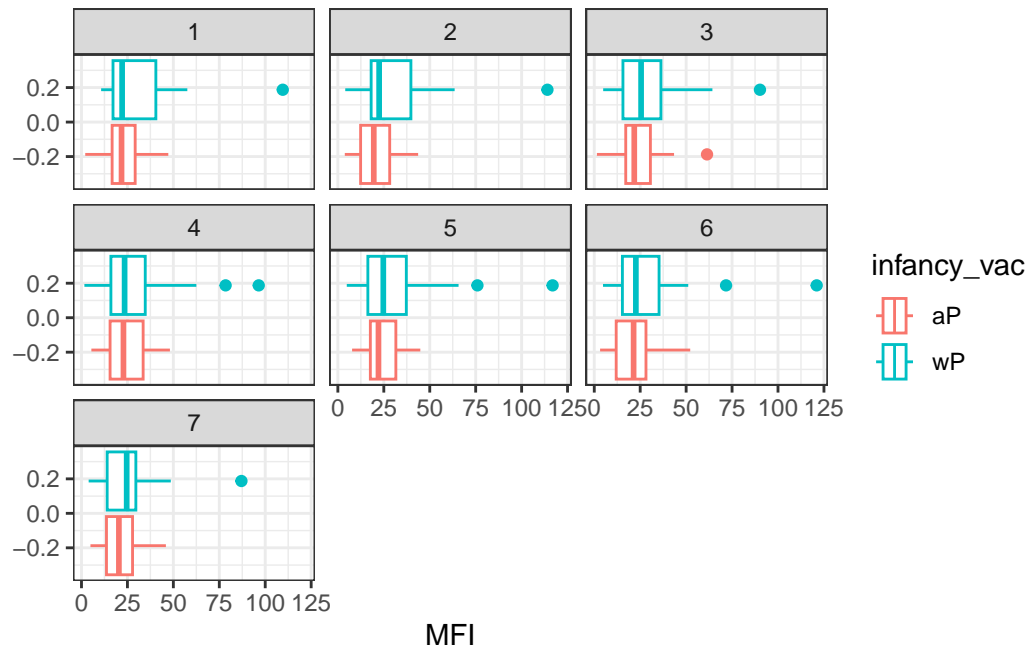


```
ggplot(ig1) +
  aes(MFI, antigen, col=infancy_vac ) +
  geom_boxplot(show.legend = FALSE) +
  facet_wrap(vars(infancy_vac, visit), nrow=2)
```

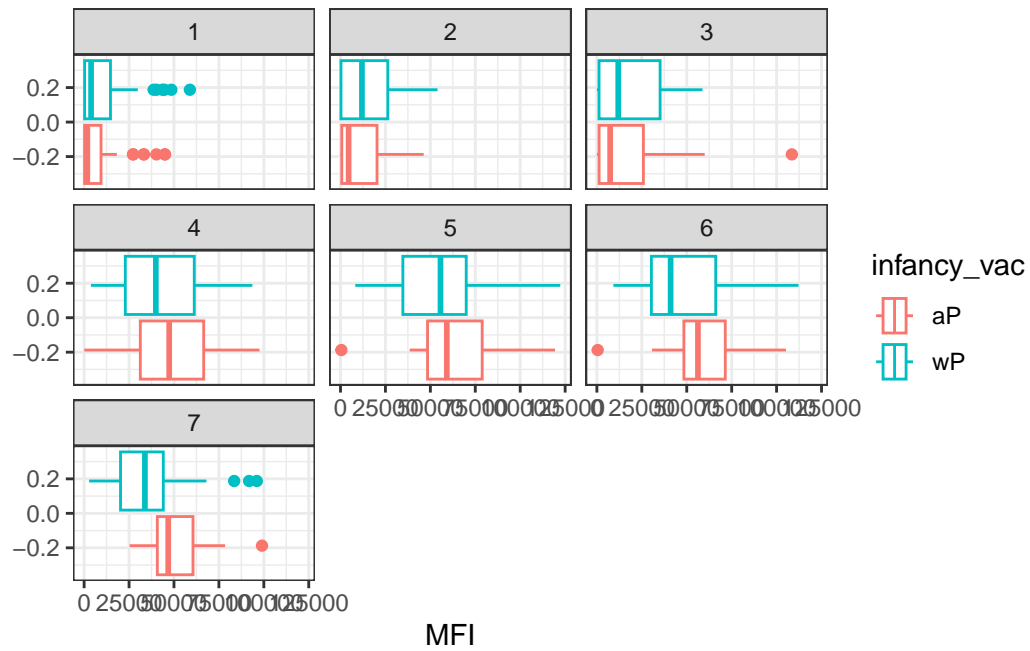


Q15. Filter to pull out only two specific antigens for analysis and create a boxplot for each. You can chose any you like. Below I picked a “control” antigen (“Measles”, that is not in our vaccines) and a clear antigen of interest (“FIM2/3”, extra-cellular fimbriae proteins from *B. pertussis* that participate in substrate attachment).

```
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()
```



FIM2/3 MFI levels rise more in comparison to Measles antigen in both aP and wP subjects.