

# Class 18: Pertussis Mini-Project

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## Background

Pertussis, also known as Whooping Cough, is a common lung infection caused by the bacteria *B. Pertussis*.

The CDC tracks cases of Pertussis in the US: <https://tinyurl.com/pertussuscde>

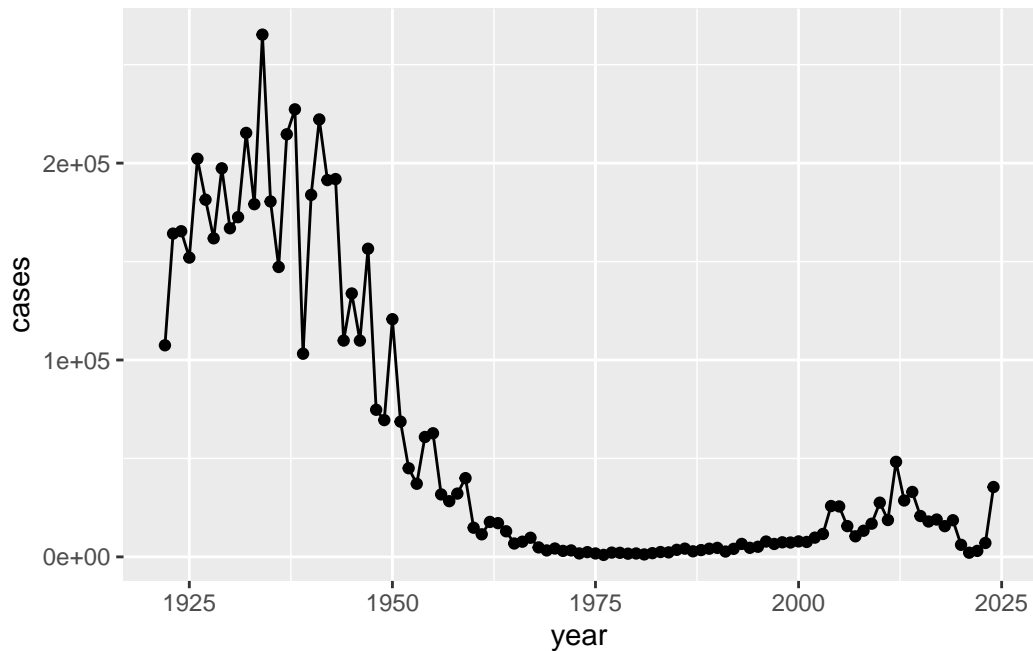
##Examine cases of Pertussis by year

We can use the **datapaste** to

Q. Make a plot of pertussis cases per year using ggplot

```
library(ggplot2)

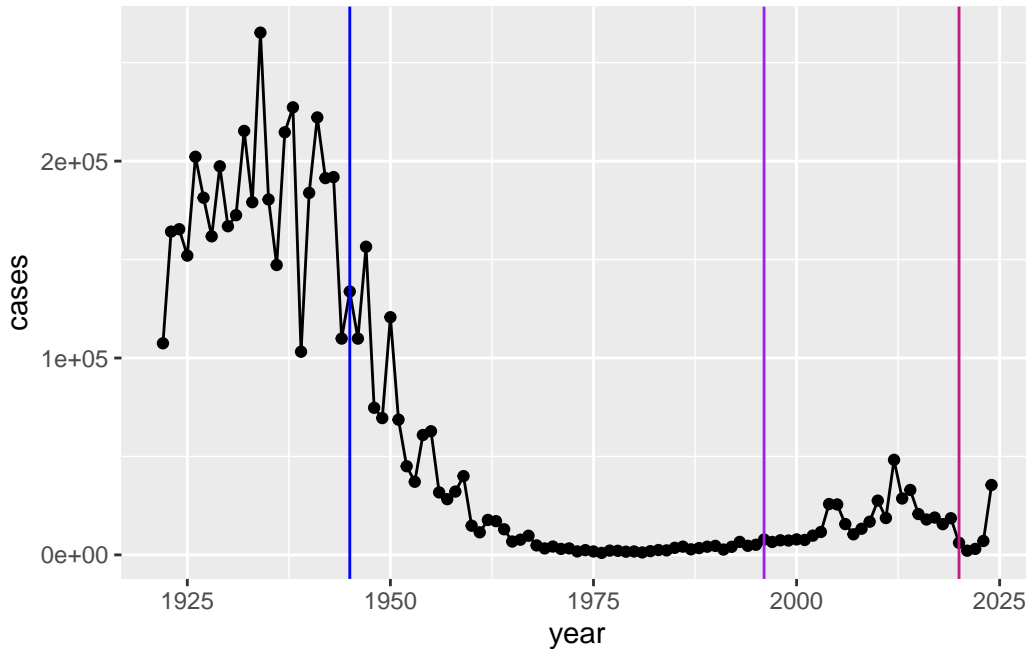
ggplot(cdc)+
  aes(year, cases)+
  geom_point()+
  geom_line()
```



Q2. Add some key time points in our history of interactions with Pertussis. These include wP roll-out (the first vaccine) in 1945 and teh swrich to aP in 1996.

We can uses `geom_vline()` which will give us a vertical line

```
ggplot(cdc)+
  aes(year, cases)+
  geom_point()+
  geom_line()+
  geom_vline(xintercept=1945, col="blue")+
  geom_vline(xintercept=1996, col="purple")+
  geom_vline(xintercept=2020, col="#C11C84")
```



Q3. Describe what happened after the introduction of the aP vaccine?

According to the graph, we can see that the aP vaccine seemed efficient as there was a decrease in the number of cases after. One of the reasons why we see any increase in the number of cases is due to anti-vaccine parents not vaccinating their children, and the virus evolving and growing resistance to the vaccine. Thailand never switched to aP vaccine and kept administering the wP vaccine and they never had a surge of cases like we see in our case.

Mounting evidence suggests that the newer **aP** vaccine is less effective over the long term than the older **wP** vaccine that it replaced. In other words, vaccine protection wanes more rapidly with the aP than the wP.

### Enter the CMI-PB Project

CMI-PB (Computational Models of Immunity- Pertussis boost) major goal is to investigate how the immune system responds differently to aP vs wP vaccinated individuals and be able to predict this at an early stage.

CMI-PB makes all their collected data freely available, and they store it in a database composed of different tables. Here we will access a few of them.

We can use the **jsonlite** package to read this 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

Q. How many subjects (i.e. enrolled people ) are in this dataset?

```
nrow(subject)
```

```
[1] 172
```

There are 172 subjects in this dataset.

Q4. How many “aP” and “wP” subjects are there

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

```
aP wP
87 85
```

There are 87 subjects who got the aP vaccine and 85 subjects got the wP vaccine.

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

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

Female	Male
112	60

There are 112 females and 60 males in this dataset

Q6. How about gender and race numbers

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

	Female	Male
American Indian/Alaska Native	0	1
Asian	32	12
Black or African American	2	3
More Than One Race	15	4
Native Hawaiian or Other Pacific Islander	1	1
Unknown or Not Reported	14	7
White	48	32

Q. Is this representative of the US population?

NOPE! But this looks like UCSD population

Lets read another database from CMI-PB

```
specimen <- read_json("https://www.cmi-pb.org/api/v5_1/specimen",  
                      simplifyVector = TRUE)  
ab_data <- read_json("https://www.cmi-pb.org/api/v5_1/plasma_ab_titer",  
                    simplifyVector = TRUE)
```

```
head(specimen)
```

	specimen_id	subject_id	actual_day_relative_to_boost
1	1	1	-3
2	2	1	1
3	3	1	3
4	4	1	7
5	5	1	11

	6	1		32
	planned_day_relative_to_boost	specimen_type	visit	
1	0	Blood	1	
2	1	Blood	2	
3	3	Blood	3	
4	7	Blood	4	
5	14	Blood	5	
6	30	Blood	6	

We want to “join” these tables together all our information together. For this we will use the **dplyr** package and the `inner_join()` function.

“

```
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	1	1	Blood
3	3	3	Blood
4	7	7	Blood
5	11	14	Blood
6	32	30	Blood

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

```
head(ab_data)
```

	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

One more “join” to get ab\_data and meta all together into one dataset

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

Joining with `by = join\_by(specimen\_id)`

```
head(abdata)
```

	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	subject_id	infancy_vac	biological_sex
1	UG/ML	2.096133	1	wP	Female
2	IU/ML	29.170000	1	wP	Female
3	IU/ML	0.530000	1	wP	Female
4	IU/ML	6.205949	1	wP	Female
5	IU/ML	4.679535	1	wP	Female
6	IU/ML	2.816431	1	wP	Female

	ethnicity	race	year_of_birth	date_of_boost	dataset
1	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
2	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
3	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
4	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
5	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
6	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset

	actual_day_relative_to_boost	planned_day_relative_to_boost	specimen_type
1	-3		Blood
2	-3		Blood
3	-3		Blood
4	-3		Blood
5	-3		Blood
6	-3		Blood

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



```
dim(abdata)
```

```
[1] 61956    20
```

Q11. How many antibody isotypes are there in the dataset?

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

```
 IgE   IgG  IgG1  IgG2  IgG3  IgG4
6698  7265 11993 12000 12000 12000
```

Q. How many different antigens are measured in the dataset?

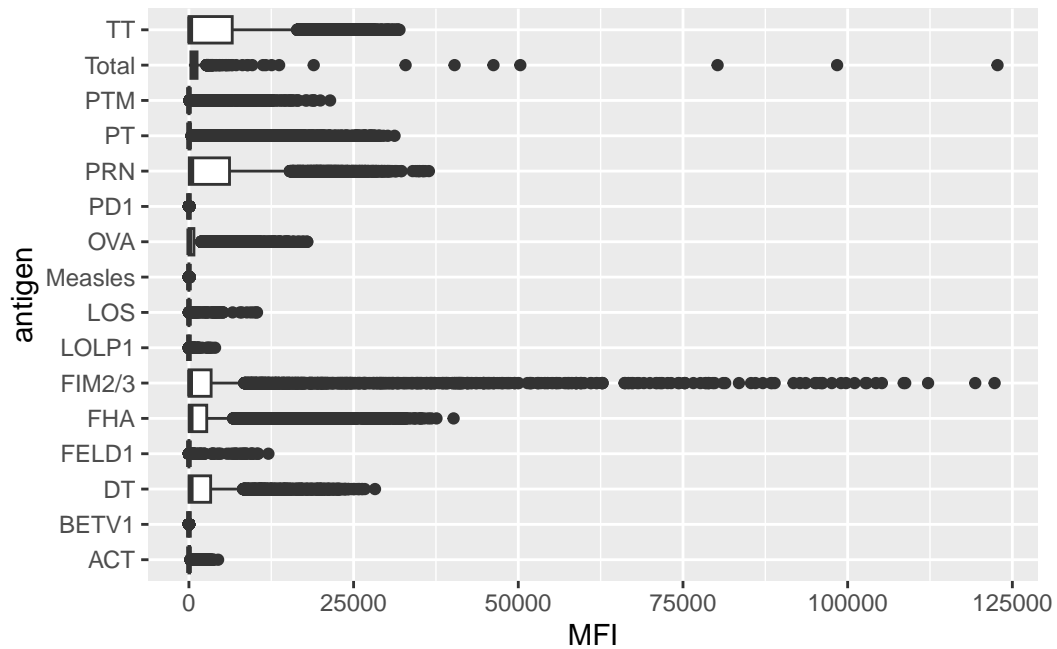
```
table(abdata$antigen)
```

```
 ACT  BETV1    DT  FELD1    FHA  FIM2/3  LOLP1    LOS Measles    OVA
1970  1970   6318  1970   6712   6318   1970   1970    1970   6318
 PD1    PRN    PT   PTM   Total    TT
1970   6712   6712  1970    788   6318
```

Q. Make a box plot of antigen levels across the whole dataset. MFI vs antigen  
MFI refers to the mean fluorescent intensity

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

```
Warning: Removed 1 row containing non-finite outside the scale range
(`stat_boxplot()`).
```

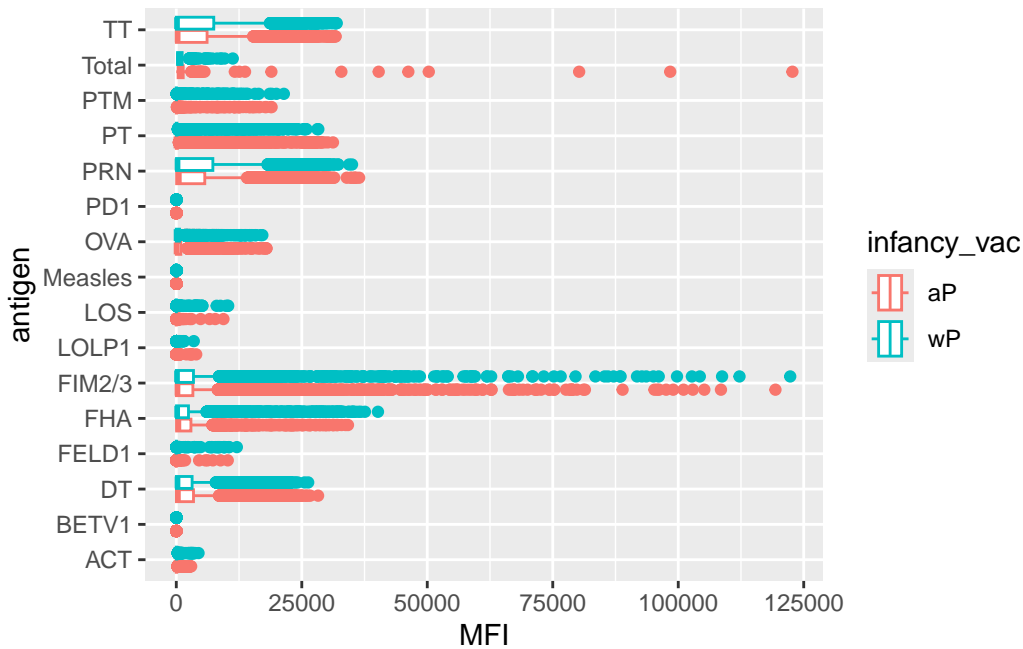


We should not see any MFI for the measles row, as that is our control

Q. Are there obvious difference between aP and wP values

```
ggplot(abdata)+
  aes(MFI, antigen, col=infancy_vac)+
  geom_boxplot()
```

Warning: Removed 1 row containing non-finite outside the scale range (`stat\_boxplot()`).



## Focus on IgG levels

IgG is the most abundant antibody in blood. With four sub-classes (IgG1 to IgG4) for crucial for long-term immunity and responding to bacterial and viral infection.

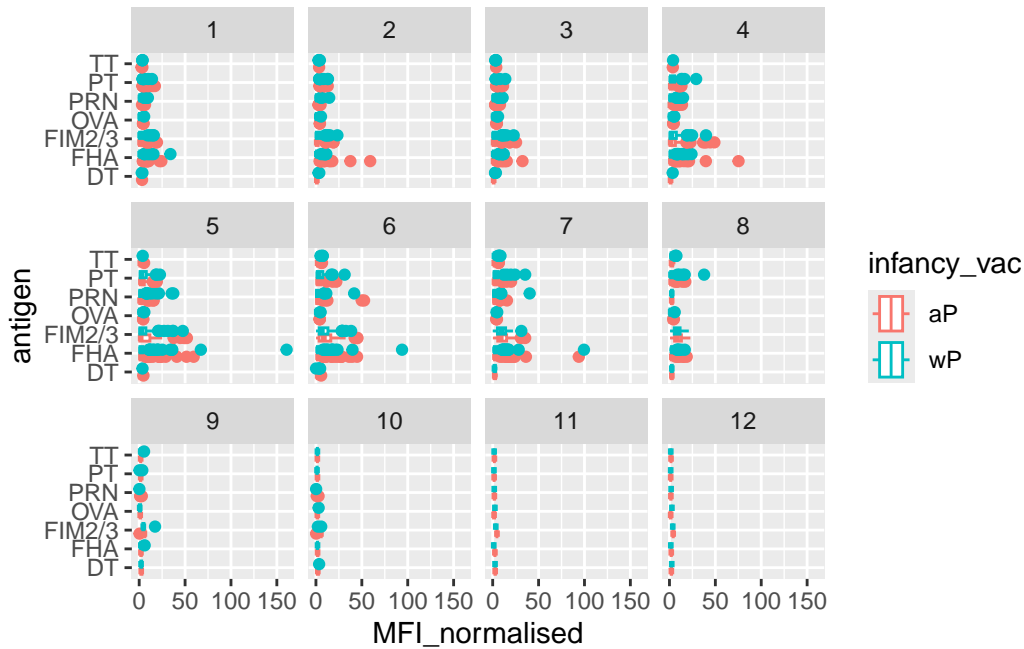
```
igg<-abdata |> filter(isotype=="IgG")
head(igg)
```

	specimen_id	isotype	is_antigen_specific	antigen	MFI	MFI_normalised
1	1	IgG	TRUE	PT	68.56614	3.736992
2	1	IgG	TRUE	PRN	332.12718	2.602350
3	1	IgG	TRUE	FHA	1887.12263	34.050956
4	19	IgG	TRUE	PT	20.11607	1.096366
5	19	IgG	TRUE	PRN	976.67419	7.652635
6	19	IgG	TRUE	FHA	60.76626	1.096457
	unit	lower_limit_of_detection	subject_id	infancy_vac	biological_sex	
1	IU/ML	0.530000	1	wP	Female	
2	IU/ML	6.205949	1	wP	Female	
3	IU/ML	4.679535	1	wP	Female	
4	IU/ML	0.530000	3	wP	Female	
5	IU/ML	6.205949	3	wP	Female	
6	IU/ML	4.679535	3	wP	Female	

	ethnicity	race	year_of_birth	date_of_boost	dataset
1	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
2	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
3	Not Hispanic or Latino	White	1986-01-01	2016-09-12	2020_dataset
4	Unknown	White	1983-01-01	2016-10-10	2020_dataset
5	Unknown	White	1983-01-01	2016-10-10	2020_dataset
6	Unknown	White	1983-01-01	2016-10-10	2020_dataset
	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				
1	1				
2	1				
3	1				
4	1				
5	1				
6	1				

Same boxplot of antigens as before

```
ggplot(igg)+
  aes(MFI_normalised, antigen, col=infancy_vac)+
  geom_boxplot()+
  facet_wrap(~visit)
```



Focus further on just one of these antigens- let's pick **PT** (Pertussis Toxin, one of the main toxins of the bacteria) in the **2021 dataset** again for the **IgG** antibody isotype.

```
table(igg$dataset)
```

```
2020_dataset 2021_dataset 2022_dataset 2023_dataset
      1182      1617      1456      3010
```

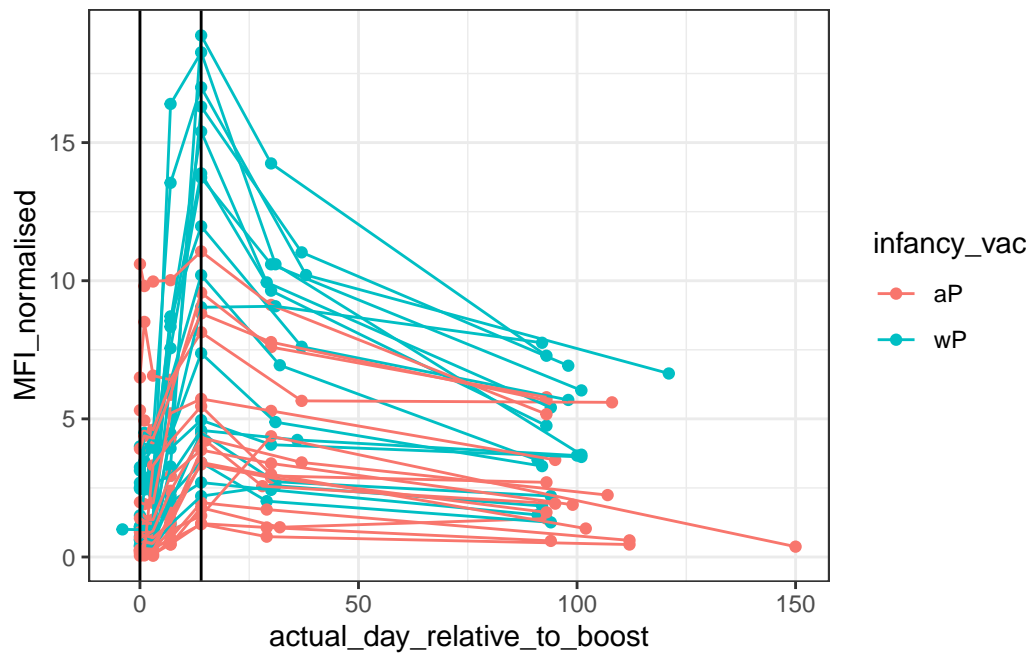
```
pt_igg<- abdata |>
  filter(isotype=="IgG", antigen=="PT", dataset=="2021_dataset")
```

```
dim(pt_igg)
```

```
[1] 231 20
```

```
ggplot(pt_igg)+
  aes(actual_day_relative_to_boost, MFI_normalised, col=infancy_vac, group=subject_id)+
  geom_point()+
  geom_line()+
  theme_bw()+
```

```
geom_vline(xintercept=0, col="black")+
geom_vline(xintercept=14, col="black")
```



At day 14, we see that we had peak levels and a distinctive difference between the aP and wP vaccines