

# Class 18: Pertussis Mini-Project

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## Background

Pertussis (aka whooping cough) is a common lung infection caused by the bacteria *B. Pertussis*.

The CDC tracks cases of Pertussis in the US: <https://www.cdc.gov/pertussis/php/surveillance/pertussis-cases-by-year.html>

## Examining cases of Pertussis by year

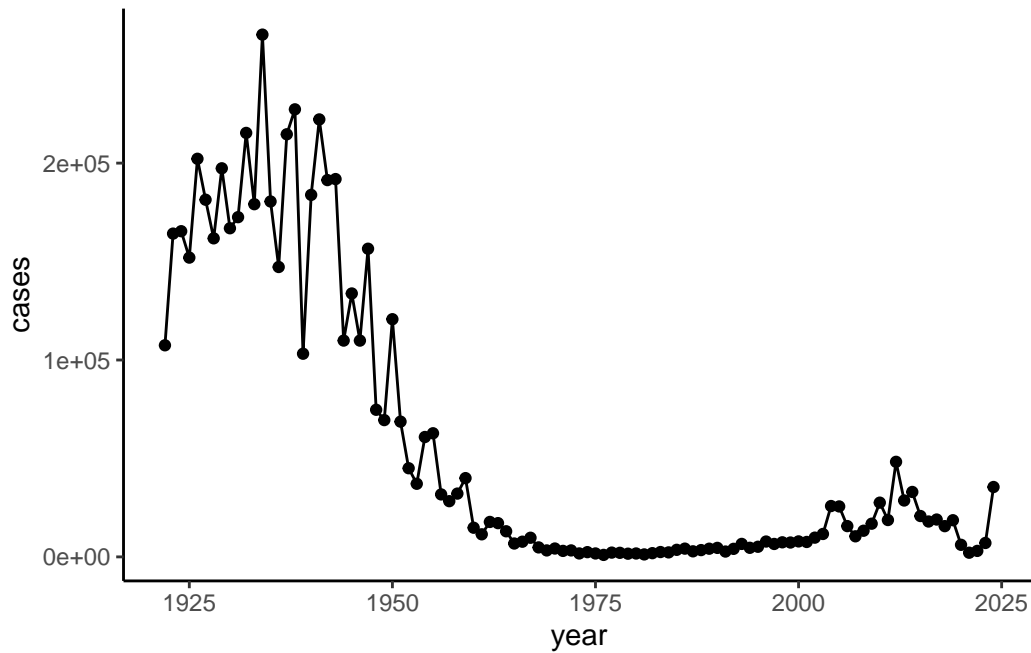
We can use the **datapasta** package to scrape case numbers from the CDC website.

Q1. Make a plot of pertussis cases per year using ggplot.

```
library(ggplot2)

cases <- ggplot(cdc) +
  aes(x=year, y=cases) +
  geom_line() +
  geom_point() +
  theme_classic()

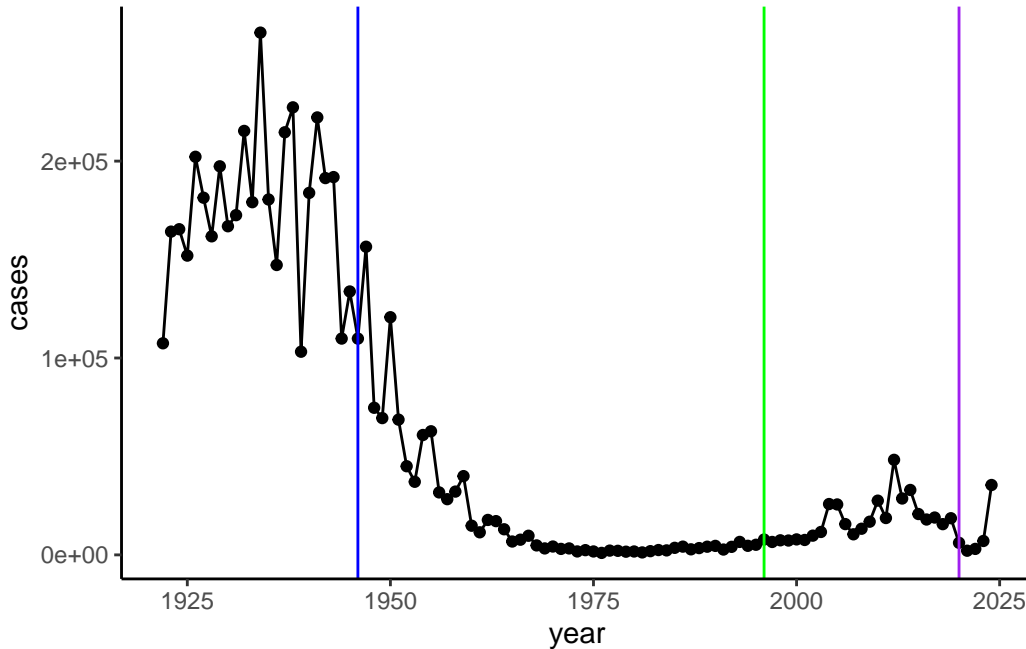
cases
```



Q2. Add some key time points in our history of interaction with Pertussis. These include wP rollout (the first vaccine) in 1946 and the switch to aP in 1996.

We can use `geom_vline()` for this.

```
cases +  
  geom_vline(xintercept = 1946, col = "blue") +  
  geom_vline(xintercept = 1996, col = "green") +  
  geom_vline(xintercept = 2020, col = "purple")
```



After the introduction of the **wP** vaccine, the number of cases decreased significantly. After the introduction of the **aP** vaccine, the number of cases remained relatively the same, but eventually increased in the years leading up to Covid (2020). The initial decrease in cases can be attributed to lockdowns and masking, but this number has since increased since national preventative orders have been lifted.

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 (efficacy) wanes more rapidly with **ap** than **wP**.

## Enter the CMI-PB project

CMI-PB (Computational Models of Immunity - Pertussis Booster shot) 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 these.

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

```
library(jsonlite)

subject <- read_json("https://www.cmi-pb.org/api/v5_1/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 there in this dataset?

```
nrow(subject)
```

```
[1] 172
```

Q. How many “aP” and “wP” subjects are there?

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

```
aP wP
87 85
```

Q. How many Male/Female are in the dataset?

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

```
Female  Male
112     60
```

Q. 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?

No, but the data is representative of the UCSD population.

Let's read another database table 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)
```

Wee peak at these:

```
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	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 to get all our information together. For this we’ll 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

We'll do one more "join" to get ab\_data and meta all together.

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

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

```
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

```
dim(abdata)
```

```
[1] 61956    20
```

Q. How many Ab 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)
```

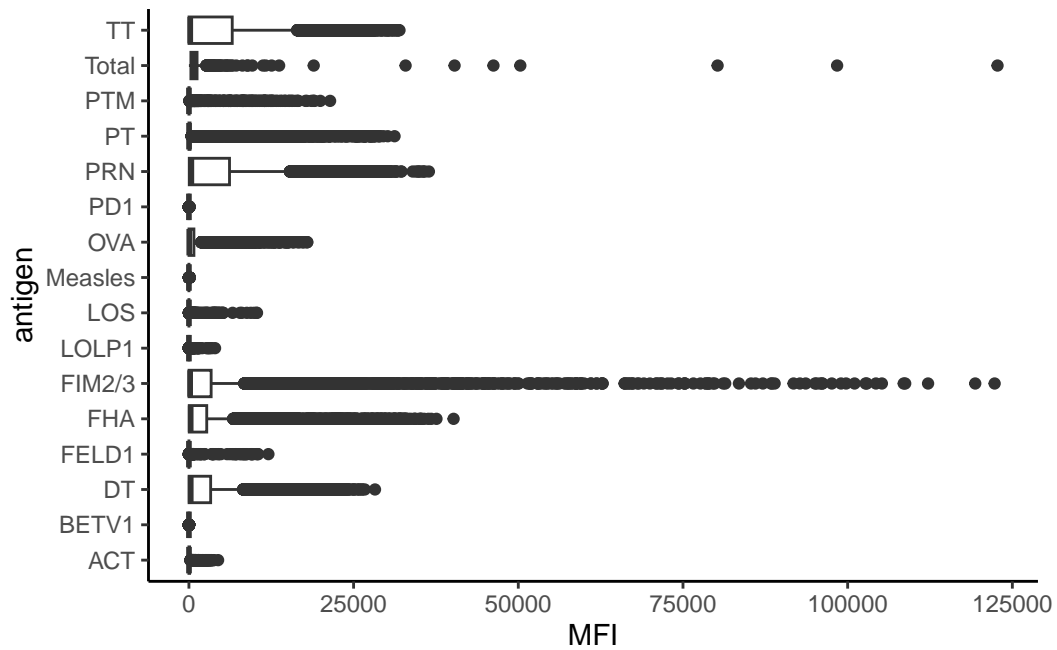
ACT	BETV1	DT	FELD1	FHA	FIM2/3	LOLP1	LOS	Measles	OVA
1970	1970	6318	1970	6712	6318	1970	1970	1970	6318
PD1	PRN	PT	PTM	Total	TT				
1970	6712	6712	1970	788	6318				

Q. Make a boxplot of antigen levels across the whole dataset (MFI vs. antigen).

```
ggplot(abdata) +
  aes(x=MFI, y=antigen) +
  geom_boxplot() +
  theme_classic()
```



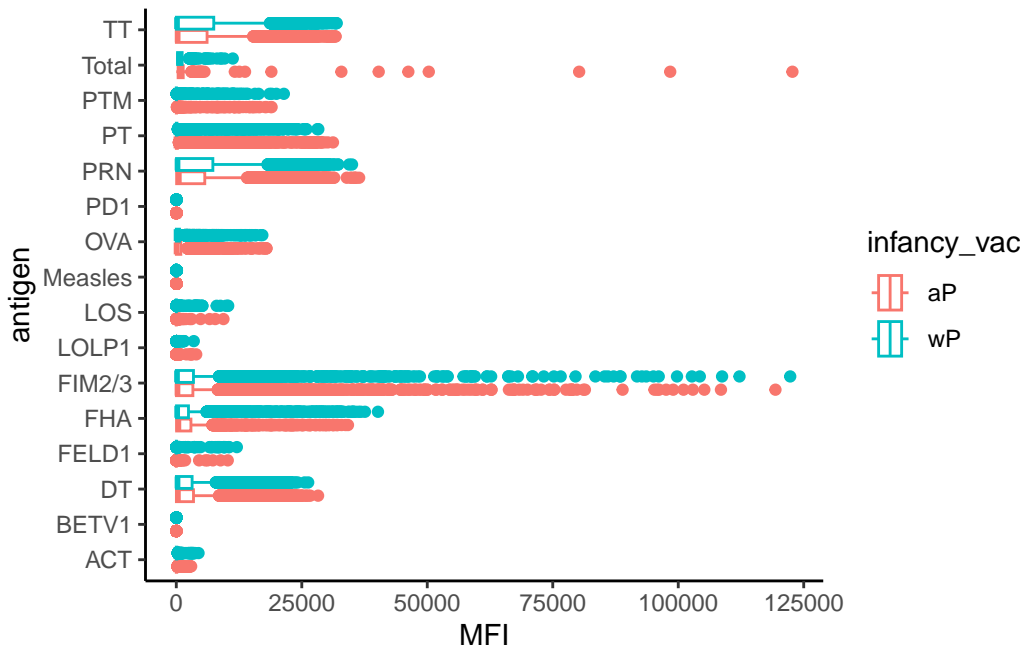
Warning: Removed 1 row containing non-finite outside the scale range  
(`stat\_boxplot()`).



Q. Are there obvious differences between aP and wP values?

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

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



## Focus on IgG levels

```
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

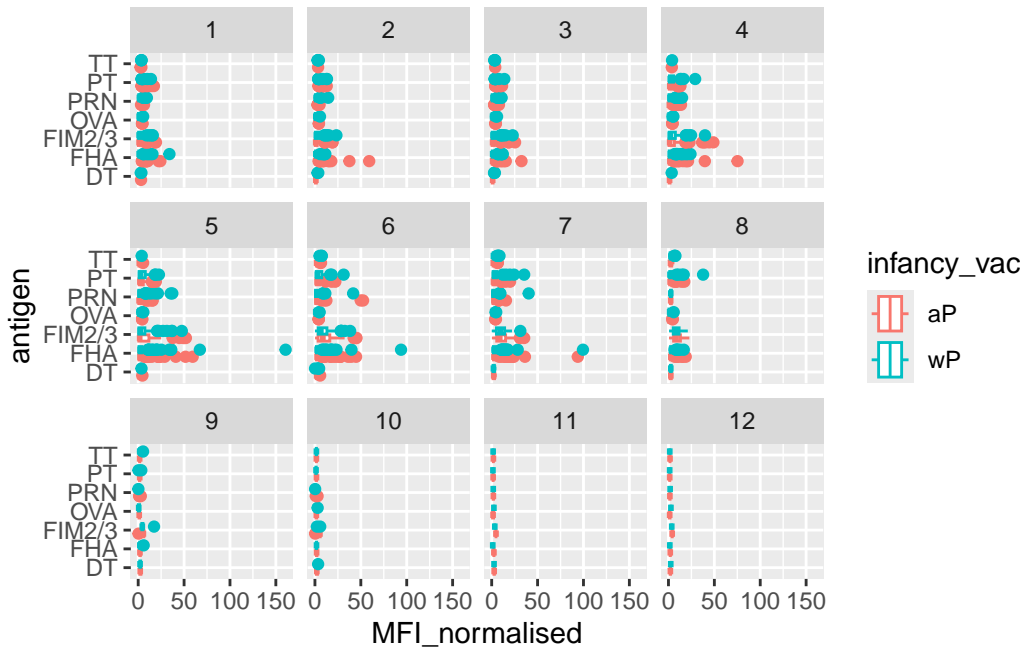
  

	ethnicity	race	year_of_birth	date_of_boost	dataset
--	-----------	------	---------------	---------------	---------

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(x=MFI_normalised, y=antigen, col=infancy_vac) +
  geom_boxplot() +
  facet_wrap(~visit)
```



Focus in 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 **IgG** antibody isotypes.

```
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")
```

```
head(pt_igg)
```

	specimen_id	isotype	is_antigen_specific	antigen	MFI	MFI_normalised	unit
1	468	IgG	FALSE	PT	112.75	1.0000000	MFI
2	469	IgG	FALSE	PT	111.25	0.9866962	MFI
3	470	IgG	FALSE	PT	125.50	1.1130820	MFI
4	471	IgG	FALSE	PT	224.25	1.9889135	MFI
5	472	IgG	FALSE	PT	304.00	2.6962306	MFI
6	473	IgG	FALSE	PT	274.00	2.4301552	MFI

	lower_limit_of_detection	subject_id	infancy_vac	biological_sex
1	5.197441	61	wP	Female
2	5.197441	61	wP	Female
3	5.197441	61	wP	Female
4	5.197441	61	wP	Female
5	5.197441	61	wP	Female
6	5.197441	61	wP	Female

	ethnicity	race	year_of_birth	date_of_boost
1	Not Hispanic or Latino Unknown or Not Reported		1987-01-01	2019-04-08
2	Not Hispanic or Latino Unknown or Not Reported		1987-01-01	2019-04-08
3	Not Hispanic or Latino Unknown or Not Reported		1987-01-01	2019-04-08
4	Not Hispanic or Latino Unknown or Not Reported		1987-01-01	2019-04-08
5	Not Hispanic or Latino Unknown or Not Reported		1987-01-01	2019-04-08
6	Not Hispanic or Latino Unknown or Not Reported		1987-01-01	2019-04-08

	dataset	actual_day_relative_to_boost	planned_day_relative_to_boost
1	2021_dataset	-4	0
2	2021_dataset	1	1
3	2021_dataset	3	3
4	2021_dataset	7	7
5	2021_dataset	14	14
6	2021_dataset	30	30

	specimen_type	visit
1	Blood	1
2	Blood	2
3	Blood	3
4	Blood	4
5	Blood	5
6	Blood	6

```
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) +
  geom_vline(xintercept=14)
```

