

# Class 15: Pertussis Mini Project

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CDC tracks Pertussis cases [here](#)

## Web Scrapping

We need to scrape this data for R. Install **datapasta**

```
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,2022L,2024L),  
  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,  
6124,2116,3044,23544)

cdc

	year	cases
1	1922	107473
2	1923	164191
3	1924	165418
4	1925	152003
5	1926	202210
6	1927	181411
7	1928	161799
8	1929	197371
9	1930	166914
10	1931	172559
11	1932	215343
12	1933	179135
13	1934	265269
14	1935	180518
15	1936	147237
16	1937	214652
17	1938	227319
18	1939	103188
19	1940	183866
20	1941	222202
21	1942	191383
22	1943	191890
23	1944	109873
24	1945	133792
25	1946	109860
26	1947	156517

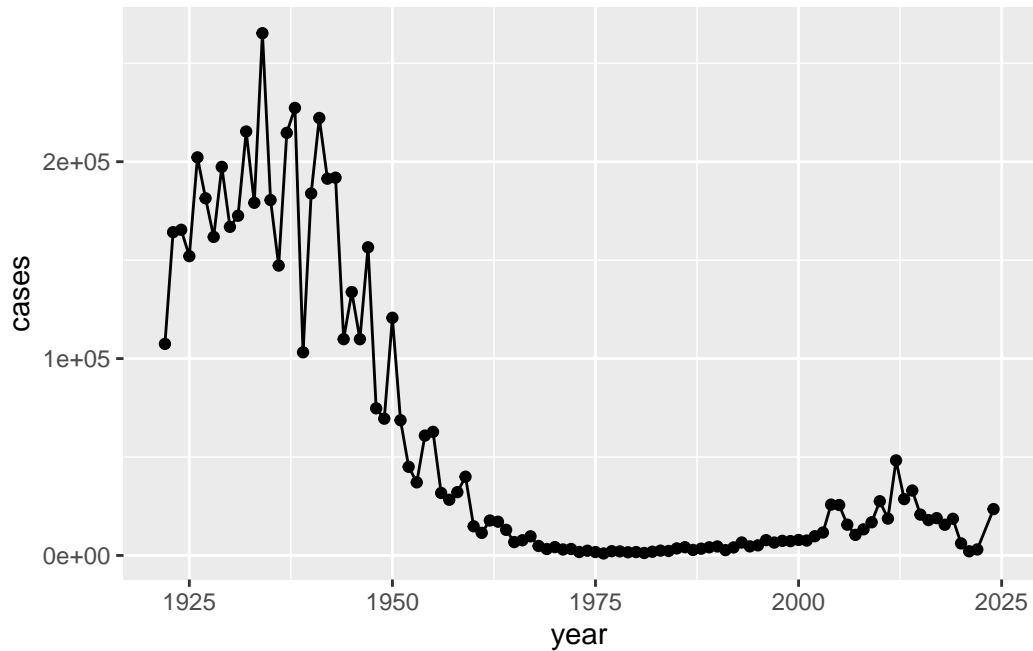
27	1948	74715
28	1949	69479
29	1950	120718
30	1951	68687
31	1952	45030
32	1953	37129
33	1954	60886
34	1955	62786
35	1956	31732
36	1957	28295
37	1958	32148
38	1959	40005
39	1960	14809
40	1961	11468
41	1962	17749
42	1963	17135
43	1964	13005
44	1965	6799
45	1966	7717
46	1967	9718
47	1968	4810
48	1969	3285
49	1970	4249
50	1971	3036
51	1972	3287
52	1973	1759
53	1974	2402
54	1975	1738
55	1976	1010
56	1977	2177
57	1978	2063
58	1979	1623
59	1980	1730
60	1981	1248
61	1982	1895
62	1983	2463
63	1984	2276
64	1985	3589
65	1986	4195
66	1987	2823
67	1988	3450
68	1989	4157
69	1990	4570

70	1991	2719
71	1992	4083
72	1993	6586
73	1994	4617
74	1995	5137
75	1996	7796
76	1997	6564
77	1998	7405
78	1999	7298
79	2000	7867
80	2001	7580
81	2002	9771
82	2003	11647
83	2004	25827
84	2005	25616
85	2006	15632
86	2007	10454
87	2008	13278
88	2009	16858
89	2010	27550
90	2011	18719
91	2012	48277
92	2013	28639
93	2014	32971
94	2015	20762
95	2016	17972
96	2017	18975
97	2018	15609
98	2019	18617
99	2020	6124
100	2021	2116
101	2022	3044
102	2024	23544

Now plot using ggplot2: Call `ggplot`, add `cdc` inside `ggplot`, add `aes`, x and y axes, and then `geom_point()` and `geom_line()`

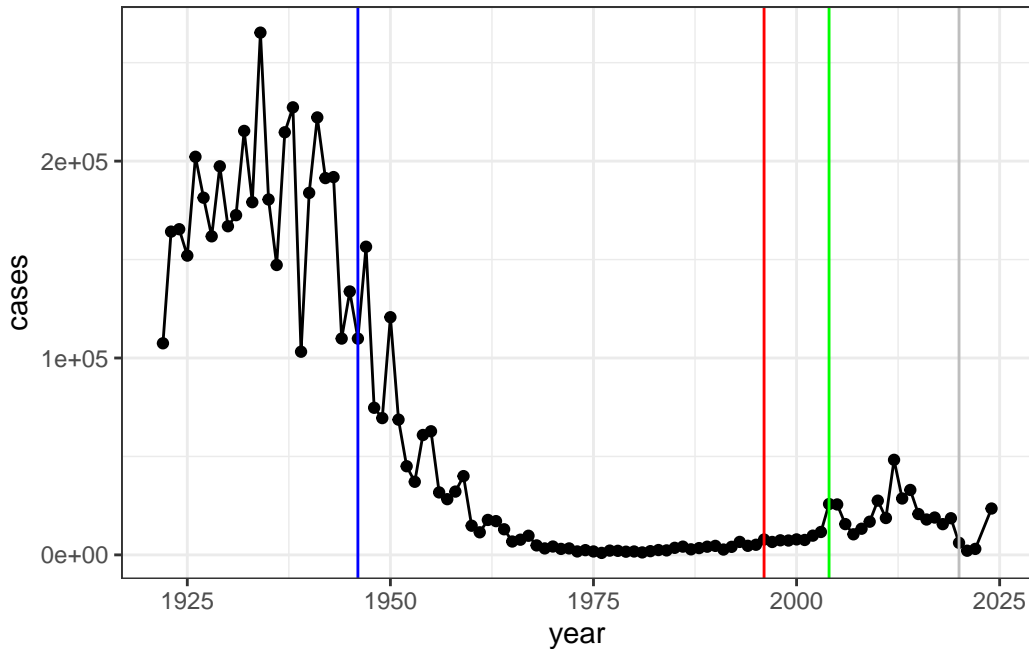
```
library(ggplot2)
baseplot<-ggplot(cdc)+
  aes(year,cases)+
  geom_point()+
  geom_line()
```

```
baseplot
```



Let's add the data of wP vaccine roll out, 1946 (blue) Let's add the data of aP vaccine roll out, 1996 (red) Now let's add 2020 and see what happened to cases that year (gray) Now, let's see the first spike after aP roll out, 2004

```
baseplot+  
  theme_bw()+  
  geom_vline(xintercept = 1946, col="blue")+  
  geom_vline(xintercept = 1996, col="red")+  
  geom_vline(xintercept = 2020, col="gray")+  
  geom_vline(xintercept = 2004, col="green")
```



## CMI-PB(Computational Models of Immunity-Pertussis Boost)

This projects collects and makes freely available data about the immune response to Pertussis vaccination Access this data via an API which returns JSON format (key:value pairs) We can use **jsonlite** package and it's `read_json()` function

```
library(jsonlite)
subject<- read_json("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
```

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

```
aP wP
87 85
```

```
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

```
table(subject$dataset)
```

```
2020_dataset 2021_dataset 2022_dataset 2023_dataset
        60             36             22             54
```

```
specimen<-read_json("specimen.json", simplifyVector = TRUE)
ab_titer<-read_json("plasma_ab_titer.json", 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	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

```
head(ab_titer)
```

	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 want to “join” these tables so we have all the info we need about a given antibody measurement

```
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

One last joining of meta and ab\_titer

```
abdata<- inner_join(ab_titer,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

```
nrow(abdata)
```

```
[1] 52576
```

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

```

IgE   IgG  IgG1  IgG2  IgG3  IgG4
6698 5389 10117 10124 10124 10124

```

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

```

      ACT  BETV1      DT  FELD1      FHA  FIM2/3  LOLP1      LOS Measles      OVA
1970    1970    4978    1970    5372    4978    1970    1970    1970    4978
      PD1      PRN      PT      PTM  Total      TT
1970    5372    5372    1970    788    4978

```

Let's begin with IgG

```

igg<- filter(abdata,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

```

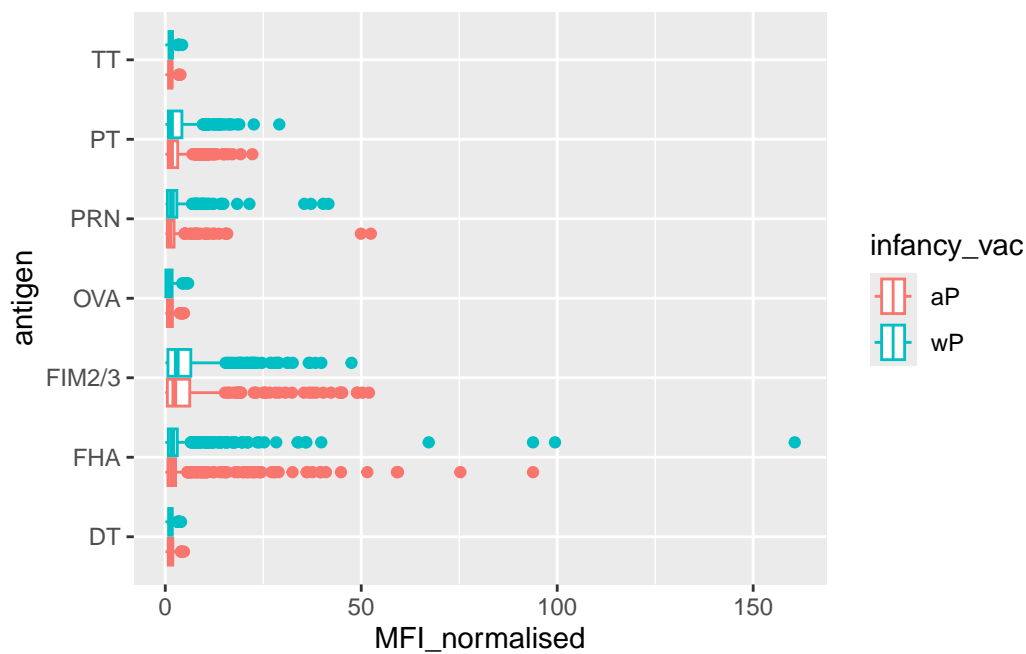
	actual_day_relative_to_boost	planned_day_relative_to_boost	specimen_type
1	-3	0	Blood
2	-3	0	Blood
3	-3	0	Blood
4	-3	0	Blood
5	-3	0	Blood
6	-3	0	Blood

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

Make a boxplot of IgG antigen levels-this will be a plot of MFI vs antigen

```
igg_boxplot<-ggplot(igg)+
  aes(MFI_normalised, antigen, col=infancy_vac)+
  geom_boxplot()
igg_boxplot
```

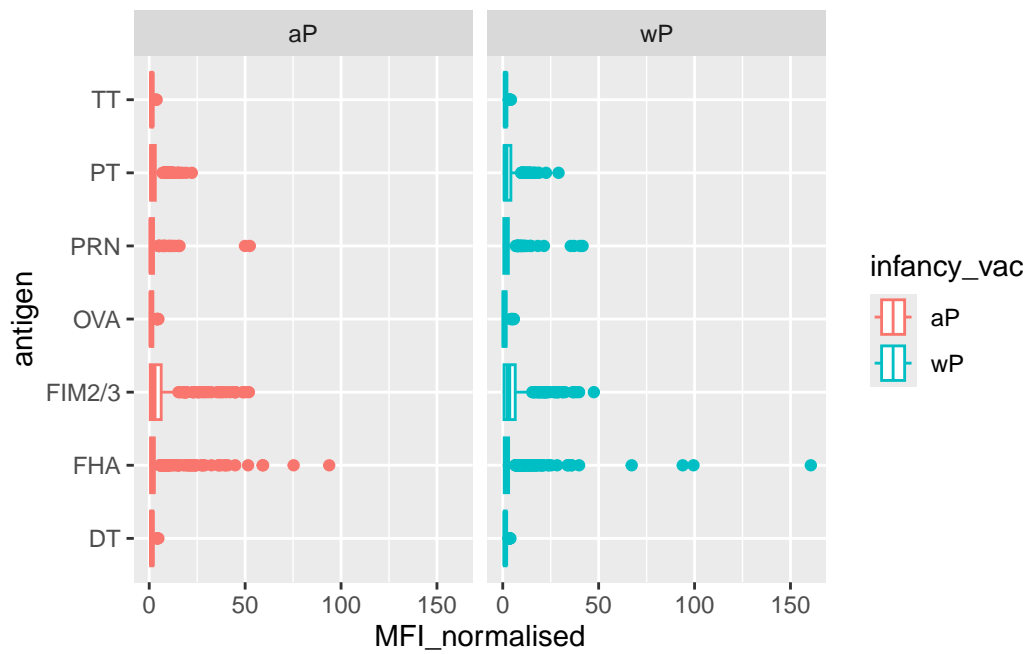


Ideally, I would like to see how these ab levels change over time relative to the booster shot

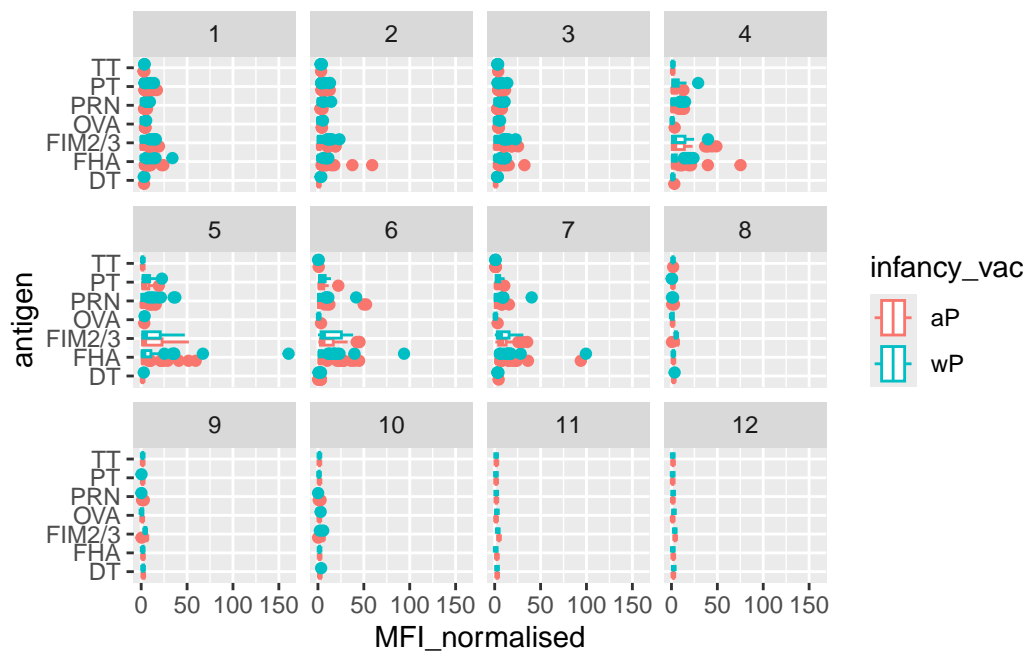
```
table(abdata$visit)
```

1	2	3	4	5	6	7	8	9	10	11	12
8280	8280	8420	6565	6565	6210	5810	815	735	686	105	105

```
igg_boxplot+
  facet_wrap(~infancy_vac)
```



```
igg_boxplot+
  facet_wrap(~visit)
```



```
abdata.21 <- abdata %>% filter(dataset == "2021_dataset")
abdata.21 %>% filter(isotype == "IgG", antigen == "PT") %>%

ggplot()+
  aes(x=planned_day_relative_to_boost, y=MFI_normalised, col=infancy_vac,group=subject_id)+
  geom_point()+
  geom_line()+
  geom_vline(xintercept=0, linetype="dashed") +
  geom_vline(xintercept=14, linetype="dashed") +
  labs(title="2021 dataset IgG PT",
        subtitle = "Dashed lines indicate day 0 (pre-boost) and 14 (apparent peak levels)")
```

## 2021 dataset IgG PT

Dashed lines indicate day 0 (pre-boost) and 14 (apparent peak levels)

