

# 5510-Projecxt

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How long did it take on average for the CFA brigade to respond to incidents in their own areas for hazard class 2 in Q4 2018/19 and Q4 2019/20? Also, this will be analyzed against the standard time (given in the data for each of the hazard classes). (CFA Emergency Response Time) . . .	1
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**How long did it take on average for the CFA brigade to respond to incidents in their own areas for hazard class 2 in Q4 2018/19 and Q4 2019/20? Also, this will be analyzed against the standard time (given in the data for each of the hazard classes). (CFA Emergency Response Time)**

## Introduction

There are three critical components of CFA delivery services: Response time: This is measured from when the brigade is alerted to an event to when the brigade arrives at the scene.

Brigade area: Each brigade has a defined brigade area, which specifies the operational footprint of the brigade.

Hazard classes: Hazard class defines the type of risk for a given area. Each brigade area may contain multiple hazard classes. However, each hazard class has a predefined service delivery standards.

In the data we have two variables that record the amount of time it CFA brigade to respond to incidents in specified areas:

- The time in which 90% of emergency incidents were responded to by any brigade; and,
- The time in which 90% of emergency incidents were responded to by the brigade within their own Brigade Area

The variable of interest for us is : The time in which 90% of emergency incidents were responded to by the brigade within their own Brigade Area. The reason is because we want to find out the mean time in which CFA brigade responded to incidents in their own brigade areas.

For this purpose, I did bit of data wrangling:

- renamed variables such that now we have a `TimeReponse` variable, and `TimeReponse_withinvariable`.
- mutate Number of incidents into a numeric vector.

- Changed the character variables to time using the contributed library chron.

For this specific question, we are going to analyze the data for **Hazard Class 2** that covers significant urban areas and is primarily residential including commercial centres, clusters of industrial and/or high density community services e.g. schools, correctional facilities, hospitals.

CFA has service delivery standards, which specify a response time target for a brigade to attend an emergency incident. The service delivery standard (response time) for Hazard Class 2 is **8 minutes**. Hence the analysis will be carried against this standard response time.

```
data_20 <- read.csv("~/5510_group_project/data/dat_q4_2020.csv")%>%
  rename(CFA_district = CFA.District)%>%
  rename(CFA_Brigade_Area = CFA.Brigade.Area)%>%
  rename(Number_of_incidents = `Number.of.emergency.incidents.within.the.Brigade.Area.for.the.reporting`)
  rename(TimeResponse = `The.time.in.which.90..of.emergency.incidents.were.responded.to.by.any.brigade`)
  rename(TimeResponse_within = `The.time.in.which.90..of.emergency.incidents.were.responded.to.by.the.brigade`)
  mutate(`TimeResponse_within` = paste0("00:",`TimeResponse_within`))%>%
  mutate(TimeResponse = paste0("00:",TimeResponse))%>%
  mutate(Number_of_incidents = as.numeric(Number_of_incidents))
```

data\_20

##	X	CFA_district	CFA_Brigade_Area	Number_of_incidents	TimeResponse
## 1	1	2	Bendigo	71	00:06:51
## 2	2	2	Eaglehawk	17	00:09:45
## 3	3	2	Golden Square	21	00:07:45
## 4	4	2	Kangaroo Flat	27	00:09:39
## 5	5	4	Portland	12	00:06:30
## 6	6	5	Warrnambool	86	00:07:28
## 7	7	7	Belmont	61	00:06:49
## 8	8	7	Corio	105	00:07:51
## 9	9	7	Geelong City	117	00:06:40
## 10	10	7	Geelong West	16	00:08:21
## 11	11	7	Lara	36	00:11:00
## 12	12	7	Ocean Grove	13	00:05:56
## 13	13	7	Torquay	36	00:11:58
## 14	14	8	Berwick	16	00:05:49
## 15	15	8	Carrum Downs	27	00:10:30
## 16	16	8	Cranbourne	91	00:08:26
## 17	17	8	Dandenong	171	00:08:17
## 18	18	8	Edithvale	13	00:07:18
## 19	19	8	Frankston	143	00:07:19
## 20	20	8	Hallam	106	00:08:10
## 21	21	8	Hampton Park	32	00:08:12
## 22	22	8	Langwarrin	17	00:10:05
## 23	23	8	Mornington	24	00:07:20
## 24	24	8	Mt Eliza	13	00:08:29
## 25	25	8	Noble Park	31	00:06:47
## 26	26	8	Pakenham	28	00:09:19
## 27	27	8	Patterson River	46	00:06:47
## 28	28	8	Rosebud	25	00:06:13
##	TimeResponse_within				
## 1					00:06:53
## 2					00:13:24

```
## 3      00:10:14
## 4      00:09:02
## 5      00:06:30
## 6      00:07:28
## 7      00:07:00
## 8      00:07:39
## 9      00:06:40
## 10     00:09:15
## 11     00:10:44
## 12     00:05:56
## 13     00:11:58
## 14     00:05:49
## 15     00:10:44
## 16     00:08:26
## 17     00:08:18
## 18     00:07:20
## 19     00:07:19
## 20     00:08:10
## 21     00:13:12
## 22     00:10:05
## 23     00:07:20
## 24     00:08:29
## 25     00:10:14
## 26     00:09:19
## 27     00:07:35
## 28     00:06:13
```

```
library(chron)
```

```
data_20$TimeResponse_within <- chron(times= data_20$TimeResponse_within)
```

```
data_20
```

```
##      X CFA_district CFA_Brigade_Area Number_of_incidents TimeResponse
## 1    1             2      Bendigo             71      00:06:51
## 2    2             2      Eaglehawk            17      00:09:45
## 3    3             2      Golden Square         21      00:07:45
## 4    4             2      Kangaroo Flat         27      00:09:39
## 5    5             4      Portland             12      00:06:30
## 6    6             5      Warrnambool           86      00:07:28
## 7    7             7      Belmont              61      00:06:49
## 8    8             7      Corio               105      00:07:51
## 9    9             7      Geelong City          117      00:06:40
## 10  10            7      Geelong West           16      00:08:21
## 11  11            7      Lara                   36      00:11:00
## 12  12            7      Ocean Grove            13      00:05:56
## 13  13            7      Torquay                36      00:11:58
## 14  14            8      Berwick                16      00:05:49
## 15  15            8      Carrum Downs           27      00:10:30
## 16  16            8      Cranbourne            91      00:08:26
## 17  17            8      Dandenong            171      00:08:17
## 18  18            8      Edithvale             13      00:07:18
## 19  19            8      Frankston            143      00:07:19
```

##	20	20	8	Hallam	106	00:08:10
##	21	21	8	Hampton Park	32	00:08:12
##	22	22	8	Langwarrin	17	00:10:05
##	23	23	8	Mornington	24	00:07:20
##	24	24	8	Mt Eliza	13	00:08:29
##	25	25	8	Noble Park	31	00:06:47
##	26	26	8	Pakenham	28	00:09:19
##	27	27	8	Patterson River	46	00:06:47
##	28	28	8	Rosebud	25	00:06:13
##	TimeResponse_within					
##	1			00:06:53		
##	2			00:13:24		
##	3			00:10:14		
##	4			00:09:02		
##	5			00:06:30		
##	6			00:07:28		
##	7			00:07:00		
##	8			00:07:39		
##	9			00:06:40		
##	10			00:09:15		
##	11			00:10:44		
##	12			00:05:56		
##	13			00:11:58		
##	14			00:05:49		
##	15			00:10:44		
##	16			00:08:26		
##	17			00:08:18		
##	18			00:07:20		
##	19			00:07:19		
##	20			00:08:10		
##	21			00:13:12		
##	22			00:10:05		
##	23			00:07:20		
##	24			00:08:29		
##	25			00:10:14		
##	26			00:09:19		
##	27			00:07:35		
##	28			00:06:13		

```
data_20$TimeResponse <- chron(times=data_20$TimeResponse)
```

```
data_20
```

##	X	CFA_district	CFA_Brigade_Area	Number_of_incidents	TimeResponse
##	1	1	2	Bendigo	71 00:06:51
##	2	2	2	Eaglehawk	17 00:09:45
##	3	3	2	Golden Square	21 00:07:45
##	4	4	2	Kangaroo Flat	27 00:09:39
##	5	5	4	Portland	12 00:06:30
##	6	6	5	Warrnambool	86 00:07:28
##	7	7	7	Belmont	61 00:06:49
##	8	8	7	Corio	105 00:07:51
##	9	9	7	Geelong City	117 00:06:40
##	10	10	7	Geelong West	16 00:08:21

## 11 11	7	Lara	36	00:11:00
## 12 12	7	Ocean Grove	13	00:05:56
## 13 13	7	Torquay	36	00:11:58
## 14 14	8	Berwick	16	00:05:49
## 15 15	8	Carrum Downs	27	00:10:30
## 16 16	8	Cranbourne	91	00:08:26
## 17 17	8	Dandenong	171	00:08:17
## 18 18	8	Edithvale	13	00:07:18
## 19 19	8	Frankston	143	00:07:19
## 20 20	8	Hallam	106	00:08:10
## 21 21	8	Hampton Park	32	00:08:12
## 22 22	8	Langwarrin	17	00:10:05
## 23 23	8	Mornington	24	00:07:20
## 24 24	8	Mt Eliza	13	00:08:29
## 25 25	8	Noble Park	31	00:06:47
## 26 26	8	Pakenham	28	00:09:19
## 27 27	8	Patterson River	46	00:06:47
## 28 28	8	Rosebud	25	00:06:13
##	TimeResponse_within			
## 1	00:06:53			
## 2	00:13:24			
## 3	00:10:14			
## 4	00:09:02			
## 5	00:06:30			
## 6	00:07:28			
## 7	00:07:00			
## 8	00:07:39			
## 9	00:06:40			
## 10	00:09:15			
## 11	00:10:44			
## 12	00:05:56			
## 13	00:11:58			
## 14	00:05:49			
## 15	00:10:44			
## 16	00:08:26			
## 17	00:08:18			
## 18	00:07:20			
## 19	00:07:19			
## 20	00:08:10			
## 21	00:13:12			
## 22	00:10:05			
## 23	00:07:20			
## 24	00:08:29			
## 25	00:10:14			
## 26	00:09:19			
## 27	00:07:35			
## 28	00:06:13			

## Context

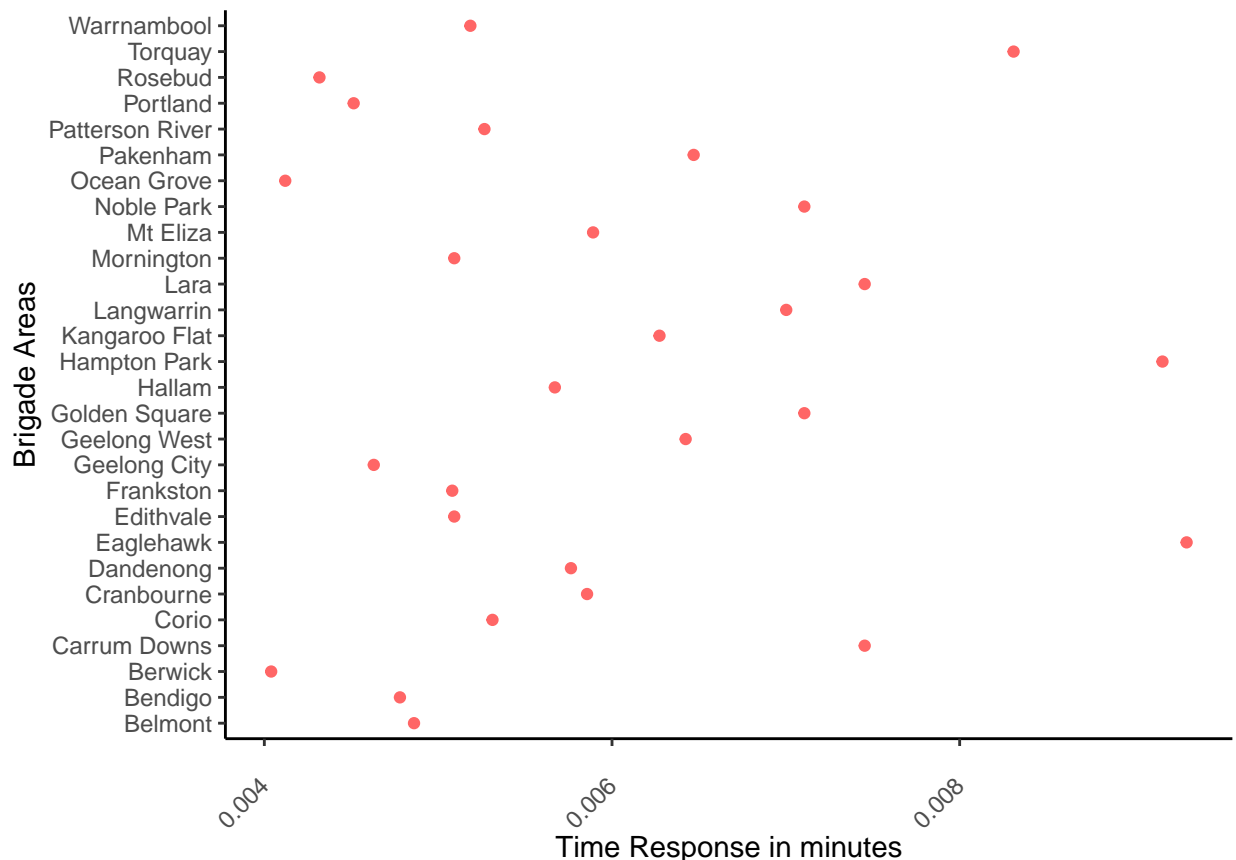
The most recent data from April to June 2020 showed CFA brigades responded to 9,120 incidents, of which 4,893 required an emergency response.

In 87 per cent of incidents that required emergency response, the community received a fire service within

the standard response time, which was broadly consistent with previous quarters.CFA News

Let's visualize the data first.

```
ggplot(data_20) +
  geom_point(data = data_20,
    aes(x = TimeResponse_within,
      y = CFA_Brigade_Area),
    color = "red",
    alpha = 0.6)+
  theme_classic()+
  theme(axis.text.x = element_text(angle = 45, vjust = 0.5, hjust=1), axis.title.x = element_text(vjust=0.5)),
  labs(x = "Time Response in minutes",
    y = "Brigade Areas")
```



```
data_20 %>%
  select(TimeResponse_within)%>%
  summary(data_20$TimeResponse_within)%>%
  kable() %>%
  kable_styling(bootstrap_options = c("striped", "hover"))
```

The mean time that it took CFA brigades to respond to incidents in their own CFA brigade areas is 0.0059838. This is very close to the target response time for brigades, which is eight minutes.

	TimeResponse_within
Min.	:00:05:49
1st Qu.	:00:07:14
Median	:00:08:14
Mean	:00:08:37
3rd Qu.	:00:10:07
Max.	:00:13:24

```
data20_avgRT <- function(TimeResponse_within){
  m = sum(TimeResponse_within)/length(TimeResponse_within)
  return(m)
}
data20_avgRT(data_20$TimeResponse_within)
```

```
## [1] 00:08:37
```

For the year 2018/19, we have used the data belonging to the year 2018/19, which was acquired from the CFA official site.

The most recent data set from July to September 2019 showed CFA brigades responding to 9,830 incidents, of which 5,013 required an emergency response.

In 87 per cent of emergencies, the community received a fire service within the standard response time, which was broadly consistent with previous quarters.CFA News

For this part of the question, I repeated the same data cleaning steps as mentioned above.

```
data_19 <- read.csv("~/5510_group_project/data/dat_q4_2018_19.csv")%>%
  rename(CFA_district = CFA.District)%>%
  rename(CFA_Brigade_Area = CFA.Brigade.Area)%>%
  rename(Number_of_incidents = `Number.of.emergency.incidents.within.the.Brigade.Area.for.the.reporting`)
  rename(TimeResponse = `The.time.in.which.90..of.emergency.incidents.were.responded.to.by.any.brigade`)
  rename(TimeResponse_within = `The.time.in.which.90..of.emergency.incidents.were.responded.to.by.the.b`)
  filter(TimeResponse_within != "NULL")%>%
  mutate(`TimeResponse_within` = paste0("00:",`TimeResponse_within`))%>%
  mutate(TimeResponse = paste0("00:",TimeResponse))%>%
  mutate(Number_of_incidents = as.numeric(Number_of_incidents))
```

```
data_19
```

```
##      X CFA_district CFA_Brigade_Area Number_of_incidents TimeResponse
## 1    1             2      Bendigo             58      00:07:32
## 2    2             2    Castlemaine             10      00:09:36
## 3    3             2    Eaglehawk              16      00:09:23
## 4    4             2  Kangaroo Flat             12      00:07:48
## 5    5             4    Portland              19      00:07:00
## 6    6             5  Warrnambool             44      00:07:13
## 7    7             6        Colac              10      00:07:46
## 8    8             7        Belmont            56      00:07:07
## 9    9             7        Corio            102      00:07:31
## 10  10            7    Geelong City            122      00:06:47
```

## 11 11	7	Lara	26	00:10:28
## 12 12	7	Ocean Grove	10	00:05:28
## 13 13	7	Torquay	15	00:11:55
## 14 14	8	Berwick	25	00:06:46
## 15 15	8	Carrum Downs	32	00:08:57
## 16 16	8	Cranbourne	78	00:09:51
## 17 17	8	Dandenong	126	00:09:13
## 18 18	8	Edithvale	13	00:06:52
## 19 19	8	Frankston	116	00:07:07
## 20 20	8	Hallam	71	00:07:23
## 21 21	8	Hampton Park	18	00:08:35
## 22 22	8	Mornington	38	00:07:07
## 23 24	8	Pakenham	27	00:08:13
## 24 25	8	Patterson River	29	00:07:33
## 25 26	8	Rosebud	17	00:07:27
## 26 27	8	Springvale	62	00:07:53
## 27 28	11	Bairnsdale	11	00:12:21
## 28 29	13	Bayswater	17	00:06:39
## 29 30	13	Boronia	46	00:06:17
## 30 31	13	Chirnside Park	14	00:08:57
## 31 32	13	Ferntree Gully	24	00:07:30
## 32 33	13	Lilydale	17	00:09:23
## 33 34	13	Montrose	16	00:06:49
## 34 35	13	Rowville	28	00:06:57
## 35 36	13	Scoresby	32	00:09:13
## 36 37	13	Warrandyte	14	00:07:27
## 37 38	14	Caroline Springs	67	00:08:57
## 38 39	14	Craigieburn	52	00:09:43
## 39 40	14	Eltham	25	00:06:48
## 40 41	14	Epping	28	00:09:09
## 41 42	14	Greenvale	17	00:08:51
## 42 43	14	Hoppers Crossing	61	00:08:18
## 43 44	14	Melton	79	00:08:04
## 44 45	14	Mernda	12	00:08:35
## 45 46	14	Point Cook	37	00:09:33
## 46 47	14	South Morang	22	00:07:24
## 47 48	14	Sunbury	44	00:08:20
## 48 49	14	Werribee	20	00:07:13
## 49 50	15	Bacchus Marsh	20	00:10:22
## 50 51	15	Ballarat	25	00:08:03
## 51 52	15	Ballarat City	58	00:05:28
## 52 53	15	Sebastopol	16	00:08:45
## 53 54	15	Wendouree	36	00:09:47
## 54 55	16	Stawell	14	00:09:29
## 55 56	18	Mildura	48	00:06:38
## 56 57	22	Shepparton	79	00:07:13
## 57 58	23	Benalla	13	00:08:34
## 58 59	23	Wangaratta	21	00:05:41
## 59 60	24	Wodonga	40	00:07:45
## 60 62	27	Morwell	39	00:06:28
## 61 63	27	Traralgon	34	00:06:34
##	TimeResponse_within			
## 1	00:07:30			
## 2	00:09:36			



## 3	00:14:45
## 4	00:07:52
## 5	00:07:00
## 6	00:07:13
## 7	00:09:02
## 8	00:08:05
## 9	00:07:56
## 10	00:06:47
## 11	00:10:28
## 12	00:05:28
## 13	00:12:59
## 14	00:06:39
## 15	00:09:21
## 16	00:09:51
## 17	00:09:13
## 18	00:07:13
## 19	00:07:18
## 20	00:08:24
## 21	00:12:16
## 22	00:07:07
## 23	00:07:51
## 24	00:08:20
## 25	00:07:27
## 26	00:08:49
## 27	00:12:23
## 28	00:06:39
## 29	00:06:27
## 30	00:10:30
## 31	00:09:45
## 32	00:09:23
## 33	00:09:07
## 34	00:07:49
## 35	00:12:51
## 36	00:10:31
## 37	00:11:32
## 38	00:09:43
## 39	00:06:48
## 40	00:10:36
## 41	00:11:31
## 42	00:08:18
## 43	00:08:04
## 44	00:08:37
## 45	00:09:33
## 46	00:07:24
## 47	00:08:20
## 48	00:10:03
## 49	00:12:11
## 50	00:10:29
## 51	00:05:27
## 52	00:11:18
## 53	00:16:38
## 54	00:09:29
## 55	00:06:38
## 56	00:07:13

```
## 57          00:08:34
## 58          00:06:15
## 59          00:07:45
## 60          00:06:28
## 61          00:06:26
```

```
library(chron)
```

```
data_19$TimeResponse_within <- chron(times= data_19$TimeResponse_within)
```

```
data_19
```

##	X	CFA_district	CFA_Brigade_Area	Number_of_incidents	TimeResponse
## 1	1	2	Bendigo	58	00:07:32
## 2	2	2	Castlemaine	10	00:09:36
## 3	3	2	Eaglehawk	16	00:09:23
## 4	4	2	Kangaroo Flat	12	00:07:48
## 5	5	4	Portland	19	00:07:00
## 6	6	5	Warrnambool	44	00:07:13
## 7	7	6	Colac	10	00:07:46
## 8	8	7	Belmont	56	00:07:07
## 9	9	7	Corio	102	00:07:31
## 10	10	7	Geelong City	122	00:06:47
## 11	11	7	Lara	26	00:10:28
## 12	12	7	Ocean Grove	10	00:05:28
## 13	13	7	Torquay	15	00:11:55
## 14	14	8	Berwick	25	00:06:46
## 15	15	8	Carrum Downs	32	00:08:57
## 16	16	8	Cranbourne	78	00:09:51
## 17	17	8	Dandenong	126	00:09:13
## 18	18	8	Edithvale	13	00:06:52
## 19	19	8	Frankston	116	00:07:07
## 20	20	8	Hallam	71	00:07:23
## 21	21	8	Hampton Park	18	00:08:35
## 22	22	8	Mornington	38	00:07:07
## 23	24	8	Pakenham	27	00:08:13
## 24	25	8	Patterson River	29	00:07:33
## 25	26	8	Rosebud	17	00:07:27
## 26	27	8	Springvale	62	00:07:53
## 27	28	11	Bairnsdale	11	00:12:21
## 28	29	13	Bayswater	17	00:06:39
## 29	30	13	Boronia	46	00:06:17
## 30	31	13	Chirnside Park	14	00:08:57
## 31	32	13	Ferntree Gully	24	00:07:30
## 32	33	13	Lilydale	17	00:09:23
## 33	34	13	Montrose	16	00:06:49
## 34	35	13	Rowville	28	00:06:57
## 35	36	13	Scoresby	32	00:09:13
## 36	37	13	Warrandyte	14	00:07:27
## 37	38	14	Caroline Springs	67	00:08:57
## 38	39	14	Craigieburn	52	00:09:43
## 39	40	14	Eltham	25	00:06:48

## 40 41	14	Epping	28	00:09:09
## 41 42	14	Greenvale	17	00:08:51
## 42 43	14	Hoppers Crossing	61	00:08:18
## 43 44	14	Melton	79	00:08:04
## 44 45	14	Mernda	12	00:08:35
## 45 46	14	Point Cook	37	00:09:33
## 46 47	14	South Morang	22	00:07:24
## 47 48	14	Sunbury	44	00:08:20
## 48 49	14	Werribee	20	00:07:13
## 49 50	15	Bacchus Marsh	20	00:10:22
## 50 51	15	Ballarat	25	00:08:03
## 51 52	15	Ballarat City	58	00:05:28
## 52 53	15	Sebastopol	16	00:08:45
## 53 54	15	Wendouree	36	00:09:47
## 54 55	16	Stawell	14	00:09:29
## 55 56	18	Mildura	48	00:06:38
## 56 57	22	Shepparton	79	00:07:13
## 57 58	23	Benalla	13	00:08:34
## 58 59	23	Wangaratta	21	00:05:41
## 59 60	24	Wodonga	40	00:07:45
## 60 62	27	Morwell	39	00:06:28
## 61 63	27	Traralgon	34	00:06:34
##	TimeResponse_within			
## 1	00:07:30			
## 2	00:09:36			
## 3	00:14:45			
## 4	00:07:52			
## 5	00:07:00			
## 6	00:07:13			
## 7	00:09:02			
## 8	00:08:05			
## 9	00:07:56			
## 10	00:06:47			
## 11	00:10:28			
## 12	00:05:28			
## 13	00:12:59			
## 14	00:06:39			
## 15	00:09:21			
## 16	00:09:51			
## 17	00:09:13			
## 18	00:07:13			
## 19	00:07:18			
## 20	00:08:24			
## 21	00:12:16			
## 22	00:07:07			
## 23	00:07:51			
## 24	00:08:20			
## 25	00:07:27			
## 26	00:08:49			
## 27	00:12:23			
## 28	00:06:39			
## 29	00:06:27			
## 30	00:10:30			
## 31	00:09:45			

```
## 32      00:09:23
## 33      00:09:07
## 34      00:07:49
## 35      00:12:51
## 36      00:10:31
## 37      00:11:32
## 38      00:09:43
## 39      00:06:48
## 40      00:10:36
## 41      00:11:31
## 42      00:08:18
## 43      00:08:04
## 44      00:08:37
## 45      00:09:33
## 46      00:07:24
## 47      00:08:20
## 48      00:10:03
## 49      00:12:11
## 50      00:10:29
## 51      00:05:27
## 52      00:11:18
## 53      00:16:38
## 54      00:09:29
## 55      00:06:38
## 56      00:07:13
## 57      00:08:34
## 58      00:06:15
## 59      00:07:45
## 60      00:06:28
## 61      00:06:26
```

```
data_19$TimeResponse <- chron(times=data_19$TimeResponse)
```

```
data_19
```

```
##      X CFA_district CFA_Brigade_Area Number_of_incidents TimeResponse
## 1    1              2      Bendigo             58      00:07:32
## 2    2              2    Castlemaine             10      00:09:36
## 3    3              2    Eaglehawk              16      00:09:23
## 4    4              2  Kangaroo Flat             12      00:07:48
## 5    5              4    Portland              19      00:07:00
## 6    6              5  Warrnambool             44      00:07:13
## 7    7              6      Colac              10      00:07:46
## 8    8              7      Belmont             56      00:07:07
## 9    9              7      Corio            102      00:07:31
## 10  10             7    Geelong City           122      00:06:47
## 11  11             7      Lara              26      00:10:28
## 12  12             7    Ocean Grove             10      00:05:28
## 13  13             7      Torquay             15      00:11:55
## 14  14             8      Berwick             25      00:06:46
## 15  15             8    Carrum Downs            32      00:08:57
## 16  16             8    Cranbourne            78      00:09:51
## 17  17             8    Dandenong           126      00:09:13
## 18  18             8    Edithvale             13      00:06:52
```

## 19 19	8	Frankston	116	00:07:07
## 20 20	8	Hallam	71	00:07:23
## 21 21	8	Hampton Park	18	00:08:35
## 22 22	8	Mornington	38	00:07:07
## 23 24	8	Pakenham	27	00:08:13
## 24 25	8	Patterson River	29	00:07:33
## 25 26	8	Rosebud	17	00:07:27
## 26 27	8	Springvale	62	00:07:53
## 27 28	11	Bairnsdale	11	00:12:21
## 28 29	13	Bayswater	17	00:06:39
## 29 30	13	Boronia	46	00:06:17
## 30 31	13	Chirnside Park	14	00:08:57
## 31 32	13	Ferntree Gully	24	00:07:30
## 32 33	13	Lilydale	17	00:09:23
## 33 34	13	Montrose	16	00:06:49
## 34 35	13	Rowville	28	00:06:57
## 35 36	13	Scoresby	32	00:09:13
## 36 37	13	Warrandyte	14	00:07:27
## 37 38	14	Caroline Springs	67	00:08:57
## 38 39	14	Craigieburn	52	00:09:43
## 39 40	14	Eltham	25	00:06:48
## 40 41	14	Epping	28	00:09:09
## 41 42	14	Greenvale	17	00:08:51
## 42 43	14	Hoppers Crossing	61	00:08:18
## 43 44	14	Melton	79	00:08:04
## 44 45	14	Mernda	12	00:08:35
## 45 46	14	Point Cook	37	00:09:33
## 46 47	14	South Morang	22	00:07:24
## 47 48	14	Sunbury	44	00:08:20
## 48 49	14	Werribee	20	00:07:13
## 49 50	15	Bacchus Marsh	20	00:10:22
## 50 51	15	Ballarat	25	00:08:03
## 51 52	15	Ballarat City	58	00:05:28
## 52 53	15	Sebastopol	16	00:08:45
## 53 54	15	Wendouree	36	00:09:47
## 54 55	16	Stawell	14	00:09:29
## 55 56	18	Mildura	48	00:06:38
## 56 57	22	Shepparton	79	00:07:13
## 57 58	23	Benalla	13	00:08:34
## 58 59	23	Wangaratta	21	00:05:41
## 59 60	24	Wodonga	40	00:07:45
## 60 62	27	Morwell	39	00:06:28
## 61 63	27	Traralgon	34	00:06:34
##	TimeResponse_within			
## 1	00:07:30			
## 2	00:09:36			
## 3	00:14:45			
## 4	00:07:52			
## 5	00:07:00			
## 6	00:07:13			
## 7	00:09:02			
## 8	00:08:05			
## 9	00:07:56			
## 10	00:06:47			

## 11	00:10:28
## 12	00:05:28
## 13	00:12:59
## 14	00:06:39
## 15	00:09:21
## 16	00:09:51
## 17	00:09:13
## 18	00:07:13
## 19	00:07:18
## 20	00:08:24
## 21	00:12:16
## 22	00:07:07
## 23	00:07:51
## 24	00:08:20
## 25	00:07:27
## 26	00:08:49
## 27	00:12:23
## 28	00:06:39
## 29	00:06:27
## 30	00:10:30
## 31	00:09:45
## 32	00:09:23
## 33	00:09:07
## 34	00:07:49
## 35	00:12:51
## 36	00:10:31
## 37	00:11:32
## 38	00:09:43
## 39	00:06:48
## 40	00:10:36
## 41	00:11:31
## 42	00:08:18
## 43	00:08:04
## 44	00:08:37
## 45	00:09:33
## 46	00:07:24
## 47	00:08:20
## 48	00:10:03
## 49	00:12:11
## 50	00:10:29
## 51	00:05:27
## 52	00:11:18
## 53	00:16:38
## 54	00:09:29
## 55	00:06:38
## 56	00:07:13
## 57	00:08:34
## 58	00:06:15
## 59	00:07:45
## 60	00:06:28
## 61	00:06:26

Let's visualize the data for the year 2018/19.



	TimeResponse_within
	Min. :00:05:27
	1st Qu.:00:07:13
	Median :00:08:24
	Mean :00:08:56
	3rd Qu.:00:10:03
	Max. :00:16:38

```
m = sum(TimeResponse_within, na.rm = TRUE)/length(TimeResponse_within)
return(m)
}
data19_avgRT(data_19$TimeResponse_within)
```

```
## [1] 00:08:56
```

## Analysis

Keeping it brief, we can state that CFA brigades' performance against standard time of eight minutes for the hazard class 2 was achieved successfully, since we see a little variation in terms of amount of time taken by the CFA.

## Which of the CFA brigade areas have had the highest number of incidents across Hazard classes for the reported period (2019/20) (four quarters). (CFA Emergency Response Time)

### Introduction

In this question, we have one variable that records the number of incidents across all hazard classes for all the specified districts. There are three predefined hazard classes: 2, 3 and 4.

- Hazard class 2 (Medium Urban): This includes significant urban areas which are primarily residential including commercial centres, clusters of industrial and/or high density community services e.g. schools, correctional facilities, hospitals. (8 minutes SDS).
- Hazard Class 3 (Low Urban): This includes all urban areas that are not included in Hazard Class 2 and includes predominantly residential occupancies and small industries. (10 minutes SDS).
- Hazard Class 4 (Rural): This includes primarily the natural surroundings in terms of bush and grassland, but also involves isolated dwellings and structures within those areas.

The variable of interest for us is : Number\_Incidents\_total.

For this purpose, I did bit of data wrangling:

- renamed variables such that now we have a CFA\_Brigade\_Area variable, and Number\_Incidents\_total variable in each data set. (I had four datasets for Quarters 1, 2, 3, and 4)
- mutate Number of incidents into a numeric vector.
- Join data using bind\_rows



## Context

In this question we are concerned with district that has had the highest number of incidents across all hazard classes in the year 2019/20.

Let's first visualize data from each data set for the quarters 1, 2, 3 and 4 for the year 2019/20.

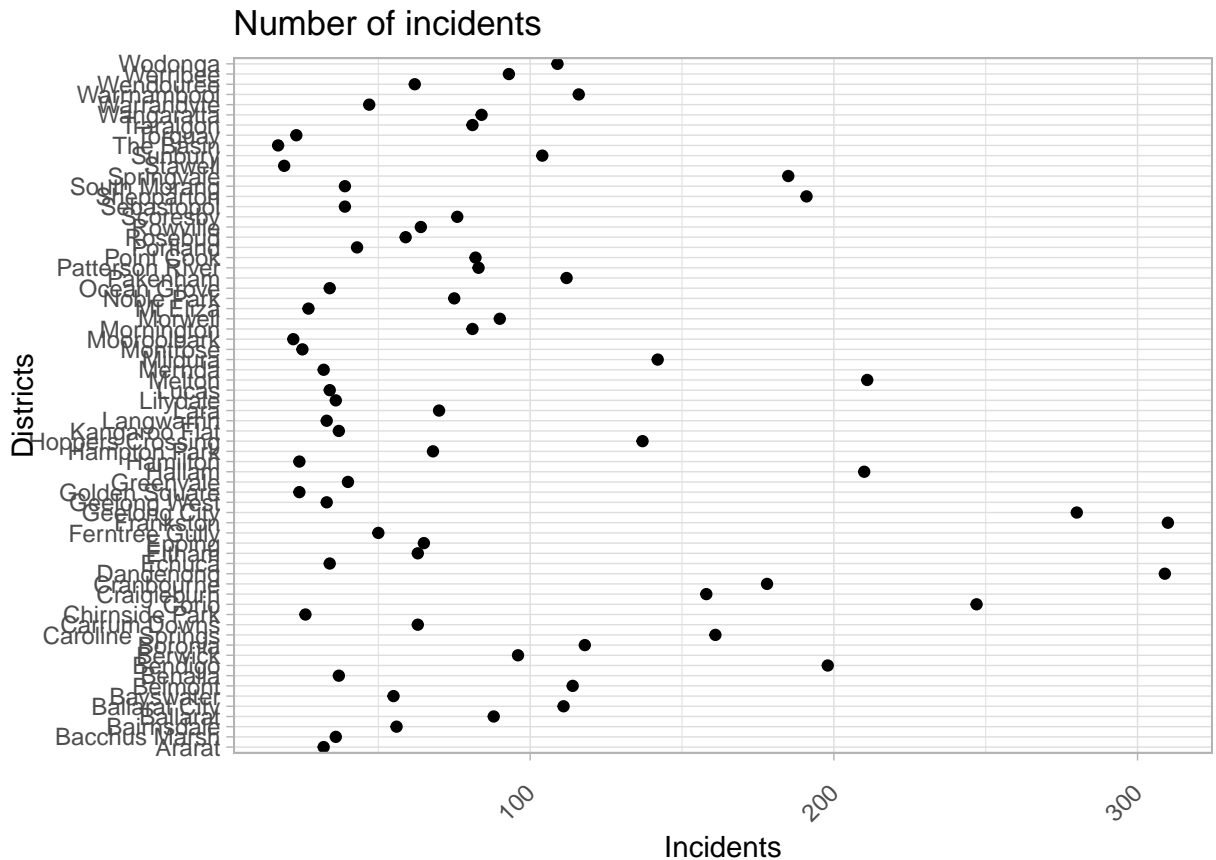
```
dat_Q1_2020 <- read_csv("~/5510_group_project/data/dat_Q1_2020.csv") %>%
  rename(Number_Incidents_total = `Number of incidents within the Brigade Area for the reporting period`)
  rename(CFA_Brigade_Area = `CFA Brigade Area`)

dat_Q1_2020
```

```
## # A tibble: 68 x 4
##       X1 'CFA District' CFA_Brigade_Area Number_Incidents_total
##   <dbl> <chr>          <chr>                  <dbl>
## 1     1 02            Bendigo                  198
## 2     2 02            Golden Square             24
## 3     3 02            Kangaroo Flat             37
## 4     4 04            Portland                  43
## 5     5 05            Hamilton                   24
## 6     6 05            Warrnambool              116
## 7     7 07            Belmont                   114
## 8     8 07            Corio                     247
## 9     9 07            Geelong City              280
## 10    10 07           Geelong West              33
## # ... with 58 more rows
```

```
ggplot(dat_Q1_2020, aes( x = Number_Incidents_total,
                        y = CFA_Brigade_Area)) +
  geom_point() +
  theme_light() +
  theme(axis.text.x = element_text(angle = 45, vjust = 0.5, hjust=1), axis.title.x = element_text(vjust=1),
        axis.text.y = element_text(angle = , vjust = 0.5, hjust=1), axis.title.y = element_text(vjust=1))

  xlab("Incidents") +
  ylab("Districts") +
  ggtitle("Number of incidents")
```



The above plot shows us the number of incidents for the first quarter 2019/20. On the x-axis we have number of incidents and the y-axis we can see the names of districts.

```
dat_Q2_2020 <- read_csv("~/5510_group_project/data/dat_Q2_2020.csv")%>%
  rename(Number_Incidents_total = `Number of incidents within the Brigade Area for the reporting period`)
  rename(CFA_Brigade_Area = `CFA Brigade Area`)
```

```
dat_Q2_2020
```

```
## # A tibble: 77 x 4
##       X1 'CFA District' CFA_Brigade_Area Number_Incidents_total
##       <dbl> <chr>          <chr>                  <dbl>
## 1      1 02             Bendigo                  225
## 2      2 02             Eaglehawk                 47
## 3      3 02             Golden Square             24
## 4      4 02             Kangaroo Flat             47
## 5      5 04             Portland                  35
## 6      6 05             Hamilton                  31
## 7      7 05             Warrnambool              124
## 8      8 06             Colac                     43
## 9      9 07             Belmont                  112
## 10     10 07            Corio                   220
## # ... with 67 more rows
```

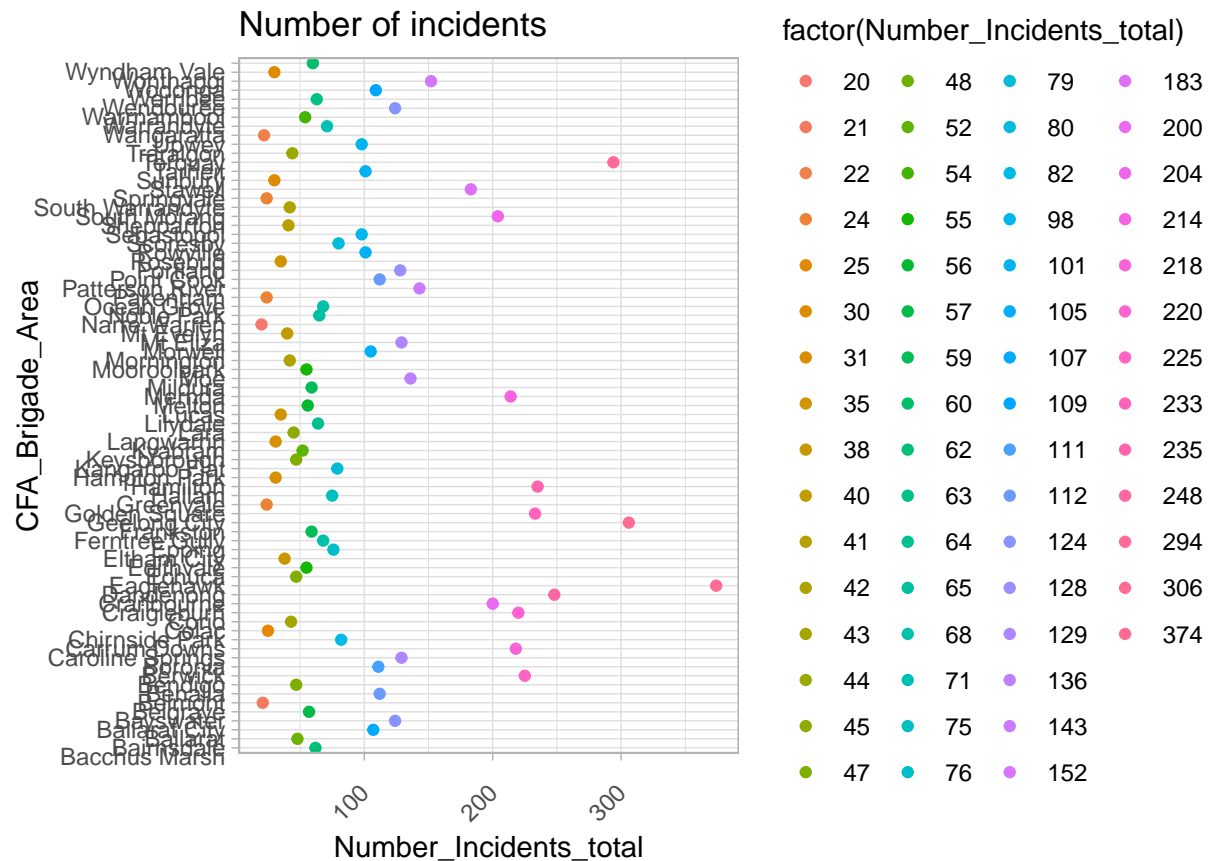
```
ggplot(dat_Q2_2020, aes(x = Number_Incidents_total ,
                        y = CFA_Brigade_Area,
```

```

colour = factor(Number_Incidents_total)))+

geom_point() +
theme_light() +
theme(axis.text.x = element_text(angle = 45, vjust = 0.5, hjust=1), axis.title.x = element_text(vjust=
theme(axis.text.y = element_text(angle = , vjust = 1, hjust=1), axis.title.y = element_text(vjust=
ggtitle("Number of incidents")

```



Similar to the plot above, this plot shows us the number of incidents for the second quarter 2019/20. On the x-axis we have number of incidents and the y-axis we can see the names of districts.

```

dat_Q3_2020 <- read_csv("~/5510_group_project/data/dat_Q3_2020.csv") %>%
  rename(Number_Incidents_total = `Number of incidents within the Brigade Area for the reporting period`)
  rename(CFA_Brigade_Area = `CFA Brigade Area`)

```

```
dat_Q3_2020
```

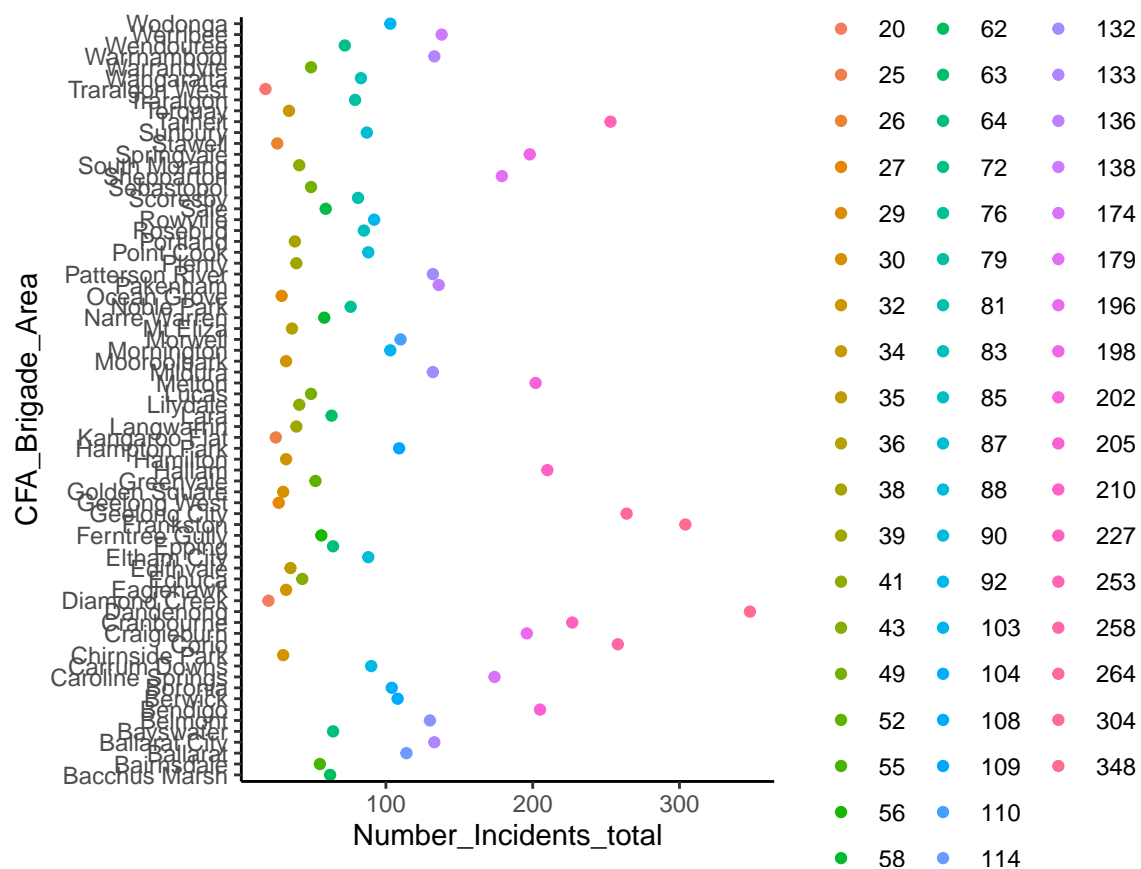
```

## # A tibble: 70 x 4
##       X1 'CFA District' CFA_Brigade_Area Number_Incidents_total
##       <dbl> <chr>          <chr>                  <dbl>
## 1       1 02            Bendigo                  205
## 2       2 02            Eaglehawk                 32
## 3       3 02            Golden Square             30
## 4       4 02            Kangaroo Flat             25

```

```
## 5      5 04      Portland      38
## 6      6 05      Hamilton      32
## 7      7 05      Warrnambool   133
## 8      8 07      Belmont      130
## 9      9 07      Corio        258
## 10     10 07     Geelong City   264
## # ... with 60 more rows
```

```
ggplot(dat_Q3_2020) +
  geom_point(aes(x = Number_Incidents_total,
  y = CFA_Brigade_Area,
  colour = factor(Number_Incidents_total))) +
  theme_classic()
```



The above plot shows us the number of incidents for the third quarter 2019/20. On the x-axis we have number of incidents and the y-axis we can see the names of districts.

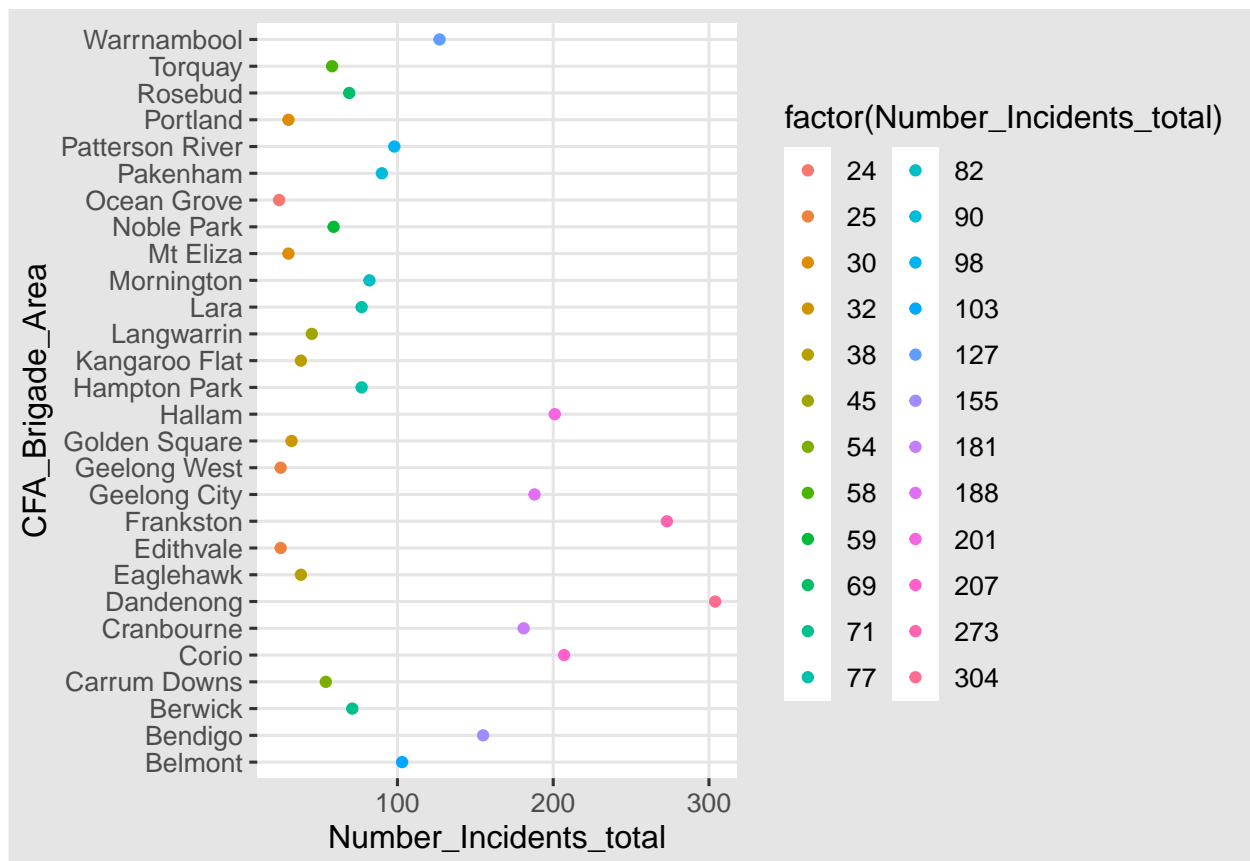
```
dat_Q4_2020 <- read_csv("~/5510_group_project/data/dat_2020Q4.csv")%>%
  select(`CFA District`, `CFA Brigade Area`, `Number of incidents within the Brigade Area for the reporting period`)
  rename(Number_Incidents_total = `Number of incidents within the Brigade Area for the reporting period`)
  rename(CFA_Brigade_Area = `CFA Brigade Area`)

dat_Q4_2020
```

```
## # A tibble: 28 x 3
```

```
##      'CFA District' CFA_Brigade_Area Number_Incidents_total
##      <chr>          <chr>                <dbl>
## 1 02              Bendigo                155
## 2 02              Eaglehawk              38
## 3 02              Golden Square          32
## 4 02              Kangaroo Flat          38
## 5 04              Portland               30
## 6 05              Warrnambool           127
## 7 07              Belmont               103
## 8 07              Corio                 207
## 9 07              Geelong City          188
## 10 07             Geelong West           25
## # ... with 18 more rows
```

```
ggplot(dat_Q4_2020) +
  geom_point(aes(x = Number_Incidents_total,
    y = CFA_Brigade_Area,
    colour = factor(Number_Incidents_total))) +
  theme_igray()
```



The above plot shows us the number of incidents for the fourth quarter 2019/20. On the x-axis we have number of incidents and the y-axis we can see the names of districts.

```
binding_data <- bind_rows(dat_Q1_2020, dat_Q2_2020, dat_Q3_2020, dat_Q4_2020)
```

```
binding_data
```

Table 1: Summary of the incidents.

	Num_incidents1	Num_incidents2	Num_incidents3	Num_incidents4
	Min. : 23.0	Min. : 24.0	Min. : 25.0	Min. : 24.0
	1st Qu.: 63.0	1st Qu.: 52.0	1st Qu.: 76.0	1st Qu.: 54.0
	Median : 83.0	Median :101.0	Median :108.0	Median : 77.0
	Mean :125.5	Mean :126.8	Mean :134.2	Mean :110.1
	3rd Qu.:185.0	3rd Qu.:183.0	3rd Qu.:198.0	3rd Qu.:181.0
	Max. :310.0	Max. :374.0	Max. :348.0	Max. :304.0

```
## # A tibble: 243 x 4
##       X1 'CFA District' CFA_Brigade_Area Number_Incidents_total
##   <dbl> <chr>          <chr>                  <dbl>
## 1     1 02          Bendigo                  198
## 2     2 02          Golden Square             24
## 3     3 02          Kangaroo Flat              37
## 4     4 04          Portland                   43
## 5     5 05          Hamilton                   24
## 6     6 05          Warrnambool               116
## 7     7 07          Belmont                   114
## 8     8 07          Corio                     247
## 9     9 07          Geelong City              280
## 10    10 07          Geelong West               33
## # ... with 233 more rows
```

## Analysis

Using the summary function we can see the maximum number of incidents across all hazard classes for specified districts for the given time frame. Data used here is the merged data of the four data sets, using bind\_rows function.

```
DataQ1_4 <- read_csv("~/5510_group_project/data/Question_3_dat_2.csv")%>%
  rename(Num_incidents1=`Number of incidents within the Brigade Area for the reporting period across all
  rename(Num_incidents2=`Number of incidents within the Brigade Area for the reporting period across all
  rename(Num_incidents3=`Number of incidents within the Brigade Area for the reporting period across all
  rename(Num_incidents4=`Number of incidents within the Brigade Area for the reporting period across all
```

```
DataQ1 <- DataQ1_4 %>%
  select(Num_incidents1,
         Num_incidents2,
         Num_incidents3,
         Num_incidents4)%>%
  summary()%>%
  kable(caption = "Summary of the incidents.") %>%
  kable_styling(bootstrap_options = c("striped", "hover"))
```

DataQ1

In the above table, we can see the summary for number of incidents in each quarter; we see that the highest number of incidents was recorded as 374 in the second quarter.

The data used here is the merged data that contains information on all four quarters. For the next steps, we will use the binding\_data from above.

Table 2: This is the summary of number of incidents across all hazard classes

Number_Incidents_total
Min. : 17.00
1st Qu.: 39.50
Median : 71.00
Mean : 95.96
3rd Qu.:125.50
Max. :374.00

Table 3: Dandenong district has the highest number of incidents across all hazard classes in all of the four quarters.

CFA_Brigade_Area	Number_Incidents_total
Dandenong	374

```
Highest_NumberIncidents <- binding_data%>%
  select(Number_Incidents_total)%>%
summary()%>%
  kable(caption = "This is the summary of number of incidents across all hazard classes") %>%
  kable_styling(bootstrap_options = c("striped", "hover"))

Highest_NumberIncidents
```

We can see that highest number of incidents are 374, so now using the filter function. let's see which district has 374 incidents.

```
binding_data%>%
  select(-X1,
    -`CFA District`)%>%
  filter(Number_Incidents_total == "374" )%>%
  kable(caption = "Dandenong district has the highest number of incidents across all hazard classes in a")
  kable_styling(bootstrap_options = c("striped", "hover"))
```

From the table we can see that Dandenong has had the highest number of incidents.

## Conclusion

As per the analysis for question on the performance of the brigade, it is safe to say that CFA brigades performed well against the standard delivery service time of eight minutes for the hazard class two incidents. And as per the analysis for the question on the number of incidents, we can see that Dandenong appears to be the area that had highest number of incidents across all hazard classes.