

hw1

AUTHOR

Andre Gala-Garza

Disclaimer: Generative AI was used to assist with templating and writing code in this assignment; however, this code was checked manually and edited by hand to ensure accuracy.

Source: OpenAI. (2026). *ChatGPT (GPT-5.2 Thinking)* [Large language model]. <https://chatgpt.com/>.

1. Exploratory data analysis

Define question

We will work with air pollution data from the U.S. Environmental Protection Agency (EPA). The EPA has a national monitoring network of air pollution sites that measure particulate matter (PM) concentrations. The primary question we will answer is whether daily concentrations of PM (particulate matter air pollution with aerodynamic diameter less than 2.5 μm) decreased in Michigan over the 20 years spanning from 2005 to 2025.

EDA Step 1: Download and read data

Please note that the [data download link](#) as given in the assignment description is currently not loading properly, so I have instead accessed the data directly.

Here is the data repository from the EPA Air Quality Data website:

https://aqs.epa.gov/aqsweb/airdata/download_files.html

The 2005 and 2025 PM_{2.5} data for all sites in Michigan were downloaded from these links:

https://aqs.epa.gov/aqsweb/airdata/daily_88101_2005.zip

https://aqs.epa.gov/aqsweb/airdata/daily_88101_2025.zip

```
pm_2005 <- read.csv("daily_88101_2005.csv")  
pm_2025 <- read.csv("daily_88101_2025.csv")
```

EDA Step 2: Check data size

```
dim(pm_2005)
```

```
[1] 145913    29
```

```
dim(pm_2025)
```

```
[1] 466760      29
```

EDA Step 3: Examine variables and their types

```
names(pm_2005)
```

```
[1] "State.Code"      "County.Code"    "Site.Num"
[4] "Parameter.Code" "POC"            "Latitude"
[7] "Longitude"       "Datum"          "Parameter.Name"
[10] "Sample.Duration" "Pollutant.Standard" "Date.Local"
[13] "Units.of.Measure" "Event.Type"      "Observation.Count"
[16] "Observation.Percent" "Arithmetic.Mean" "X1st.Max.Value"
[19] "X1st.Max.Hour"    "AQI"            "Method.Code"
[22] "Method.Name"      "Local.Site.Name" "Address"
[25] "State.Name"       "County.Name"    "City.Name"
[28] "CBSA.Name"        "Date.of.Last.Change"
```

```
names(pm_2025)
```

```
[1] "State.Code"      "County.Code"    "Site.Num"
[4] "Parameter.Code" "POC"            "Latitude"
[7] "Longitude"       "Datum"          "Parameter.Name"
[10] "Sample.Duration" "Pollutant.Standard" "Date.Local"
[13] "Units.of.Measure" "Event.Type"      "Observation.Count"
[16] "Observation.Percent" "Arithmetic.Mean" "X1st.Max.Value"
[19] "X1st.Max.Hour"    "AQI"            "Method.Code"
[22] "Method.Name"      "Local.Site.Name" "Address"
[25] "State.Name"       "County.Name"    "City.Name"
[28] "CBSA.Name"        "Date.of.Last.Change"
```

```
str(pm_2005)
```

```
'data.frame':  145913 obs. of  29 variables:
 $ State.Code      : chr  "01" "01" "01" "01" ...
 $ County.Code     : int   3 3 3 3 3 3 3 3 3 3 ...
 $ Site.Num        : int   10 10 10 10 10 10 10 10 10 10 ...
 $ Parameter.Code  : int  88101 88101 88101 88101 88101 88101 88101 88101 88101 88101 ...
 $ POC             : int   1 1 1 1 1 1 1 1 1 1 ...
 $ Latitude        : num   30.5 30.5 30.5 30.5 30.5 ...
 $ Longitude       : num  -87.9 -87.9 -87.9 -87.9 -87.9 ...
 $ Datum           : chr   "NAD83" "NAD83" "NAD83" "NAD83" ...
 $ Parameter.Name  : chr   "PM2.5 - Local Conditions" "PM2.5 - Local Conditions" "PM2.5 - Local
Conditions" "PM2.5 - Local Conditions" ...
 $ Sample.Duration : chr   "24 HOUR" "24 HOUR" "24 HOUR" "24 HOUR" ...
 $ Pollutant.Standard : chr  "PM25 24-hour 2012" "PM25 24-hour 2012" "PM25 24-hour 2012" "PM25
```

24-hour 2012" ...

\$ Date.Local : chr "2005-01-01" "2005-01-04" "2005-01-07" "2005-01-10" ...

\$ Units.of.Measure : chr "Micrograms/cubic meter (LC)" "Micrograms/cubic meter (LC)"

"Micrograms/cubic meter (LC)" "Micrograms/cubic meter (LC)" ...

\$ Event.Type : chr "None" "None" "None" "None" ...

\$ Observation.Count : int 1 1 1 1 1 1 1 1 1 1 ...

\$ Observation.Percent: num 100 100 100 100 100 100 100 100 100 100 ...

\$ Arithmetic.Mean : num 9 7.7 7.3 8.7 3.3 7.7 16.6 7.9 15.1 23.4 ...

\$ X1st.Max.Value : num 9 7.7 7.3 8.7 3.3 7.7 16.6 7.9 15.1 23.4 ...

\$ X1st.Max.Hour : int 0 0 0 0 0 0 0 0 0 0 ...

\$ AQI : int 50 43 41 48 18 43 65 44 62 78 ...

\$ Method.Code : int 118 118 118 118 118 118 118 118 118 118 ...

\$ Method.Name : chr "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" ...

\$ Local.Site.Name : chr "FAIRHOPE, Alabama" "FAIRHOPE, Alabama" "FAIRHOPE, Alabama" "FAIRHOPE, Alabama" ...

\$ Address : chr "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" ...

\$ State.Name : chr "Alabama" "Alabama" "Alabama" "Alabama" ...

\$ County.Name : chr "Baldwin" "Baldwin" "Baldwin" "Baldwin" ...

\$ City.Name : chr "Fairhope" "Fairhope" "Fairhope" "Fairhope" ...

\$ CBSA.Name : chr "Daphne-Fairhope-Foley, AL" "Daphne-Fairhope-Foley, AL" "Daphne-Fairhope-Foley, AL" "Daphne-Fairhope-Foley, AL" ...

\$ Date.of.Last.Change: chr "2024-09-05" "2024-09-05" "2024-09-05" "2024-09-05" ...

`str(pm_2025)`

'data.frame': 466760 obs. of 29 variables:

\$ State.Code : int 1 1 1 1 1 1 1 1 1 1 ...

\$ County.Code : int 3 3 3 3 3 3 3 3 3 3 ...

\$ Site.Num : int 10 10 10 10 10 10 10 10 10 10 ...

\$ Parameter.Code : int 88101 88101 88101 88101 88101 88101 88101 88101 88101 88101 ...

\$ POC : int 3 3 3 3 3 3 3 3 3 3 ...

\$ Latitude : num 30.5 30.5 30.5 30.5 30.5 ...

\$ Longitude : num -87.9 -87.9 -87.9 -87.9 -87.9 ...

\$ Datum : chr "NAD83" "NAD83" "NAD83" "NAD83" ...

\$ Parameter.Name : chr "PM2.5 - Local Conditions" "PM2.5 - Local Conditions" "PM2.5 - Local Conditions" "PM2.5 - Local Conditions" ...

\$ Sample.Duration : chr "1 HOUR" "1 HOUR" "1 HOUR" "1 HOUR" ...

\$ Pollutant.Standard : chr "" "" "" "" ...

\$ Date.Local : chr "2025-01-01" "2025-01-02" "2025-01-03" "2025-01-04" ...

\$ Units.of.Measure : chr "Micrograms/cubic meter (LC)" "Micrograms/cubic meter (LC)"

"Micrograms/cubic meter (LC)" "Micrograms/cubic meter (LC)" ...

\$ Event.Type : chr "None" "None" "None" "None" ...

\$ Observation.Count : int 24 24 24 24 24 23 24 24 24 24 ...

\$ Observation.Percent: num 100 100 100 100 100 96 100 100 100 100 ...

\$ Arithmetic.Mean : num 3.62 6.79 9.96 5.5 5.5 ...

\$ X1st.Max.Value : num 19 26 25 11 12 16 5 14 35 16 ...

```

$ X1st.Max.Hour      : int   21 23 1 19 14 2 9 19 22 0 ...
$ AQI                : int   NA NA NA NA NA NA NA NA NA ...
$ Method.Code        : int   209 209 209 209 209 209 209 209 209 ...
$ Method.Name        : chr   "Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta
Attenuation" "Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation" "Met One
BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation" "Met One BAM-1022 Mass Monitor w/
VSCC or TE-PM2.5C - Beta Attenuation" ...
$ Local.Site.Name    : chr   "FAIRHOPE, Alabama" "FAIRHOPE, Alabama" "FAIRHOPE, Alabama"
"FAIRHOPE, Alabama" ...
$ Address            : chr   "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" "FAIRHOPE
HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE,
ALABAMA" "FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA" ...
$ State.Name         : chr   "Alabama" "Alabama" "Alabama" "Alabama" ...
$ County.Name        : chr   "Baldwin" "Baldwin" "Baldwin" "Baldwin" ...
$ City.Name          : chr   "Fairhope" "Fairhope" "Fairhope" "Fairhope" ...
$ CBSA.Name          : chr   "Daphne-Fairhope-Foley, AL" "Daphne-Fairhope-Foley, AL" "Daphne-
Fairhope-Foley, AL" "Daphne-Fairhope-Foley, AL" ...
$ Date.of.Last.Change: chr   "2025-10-29" "2025-10-29" "2025-10-29" "2025-10-29" ...

```

EDA Step 4: Look at top/bottom of data

```
head(pm_2005)
```

```

State.Code County.Code Site.Num Parameter.Code POC Latitude Longitude Datum
1          01          3       10          88101  1 30.49748 -87.88026 NAD83
2          01          3       10          88101  1 30.49748 -87.88026 NAD83
3          01          3       10          88101  1 30.49748 -87.88026 NAD83
4          01          3       10          88101  1 30.49748 -87.88026 NAD83
5          01          3       10          88101  1 30.49748 -87.88026 NAD83
6          01          3       10          88101  1 30.49748 -87.88026 NAD83

Parameter.Name Sample.Duration Pollutant.Standard Date.Local
1 PM2.5 - Local Conditions      24 HOUR PM25 24-hour 2012 2005-01-01
2 PM2.5 - Local Conditions      24 HOUR PM25 24-hour 2012 2005-01-04
3 PM2.5 - Local Conditions      24 HOUR PM25 24-hour 2012 2005-01-07
4 PM2.5 - Local Conditions      24 HOUR PM25 24-hour 2012 2005-01-10
5 PM2.5 - Local Conditions      24 HOUR PM25 24-hour 2012 2005-01-13
6 PM2.5 - Local Conditions      24 HOUR PM25 24-hour 2012 2005-01-16

Units.of.Measure Event.Type Observation.Count Observation.Percent
1 Micrograms/cubic meter (LC)      None              1              100
2 Micrograms/cubic meter (LC)      None              1              100
3 Micrograms/cubic meter (LC)      None              1              100
4 Micrograms/cubic meter (LC)      None              1              100
5 Micrograms/cubic meter (LC)      None              1              100
6 Micrograms/cubic meter (LC)      None              1              100

Arithmetic.Mean X1st.Max.Value X1st.Max.Hour AQI Method.Code
1              9.0              9.0              0 50          118
2              7.7              7.7              0 43          118
3              7.3              7.3              0 41          118
4              8.7              8.7              0 48          118

```

5		3.3		3.3		0	18	118
6		7.7		7.7		0	43	118
							Method.Name	Local.Site.Name
1	R & P Model	2025	PM2.5	Sequential	w/WINS	-	GRAVIMETRIC	FAIRHOPE, Alabama
2	R & P Model	2025	PM2.5	Sequential	w/WINS	-	GRAVIMETRIC	FAIRHOPE, Alabama
3	R & P Model	2025	PM2.5	Sequential	w/WINS	-	GRAVIMETRIC	FAIRHOPE, Alabama
4	R & P Model	2025	PM2.5	Sequential	w/WINS	-	GRAVIMETRIC	FAIRHOPE, Alabama
5	R & P Model	2025	PM2.5	Sequential	w/WINS	-	GRAVIMETRIC	FAIRHOPE, Alabama
6	R & P Model	2025	PM2.5	Sequential	w/WINS	-	GRAVIMETRIC	FAIRHOPE, Alabama
							Address	State.Name
1	FAIRHOPE HIGH SCHOOL,	1	PIRATE DRIVE,	FAIRHOPE,	ALABAMA		Alabama	
2	FAIRHOPE HIGH SCHOOL,	1	PIRATE DRIVE,	FAIRHOPE,	ALABAMA		Alabama	
3	FAIRHOPE HIGH SCHOOL,	1	PIRATE DRIVE,	FAIRHOPE,	ALABAMA		Alabama	
4	FAIRHOPE HIGH SCHOOL,	1	PIRATE DRIVE,	FAIRHOPE,	ALABAMA		Alabama	
5	FAIRHOPE HIGH SCHOOL,	1	PIRATE DRIVE,	FAIRHOPE,	ALABAMA		Alabama	
6	FAIRHOPE HIGH SCHOOL,	1	PIRATE DRIVE,	FAIRHOPE,	ALABAMA		Alabama	
County.Name		City.Name		CBSA.Name		Date.of.Last.Change		
1	Baldwin	Fairhope	Daphne-Fairhope-Foley, AL				2024-09-05	
2	Baldwin	Fairhope	Daphne-Fairhope-Foley, AL				2024-09-05	
3	Baldwin	Fairhope	Daphne-Fairhope-Foley, AL				2024-09-05	
4	Baldwin	Fairhope	Daphne-Fairhope-Foley, AL				2024-09-05	
5	Baldwin	Fairhope	Daphne-Fairhope-Foley, AL				2024-09-05	
6	Baldwin	Fairhope	Daphne-Fairhope-Foley, AL				2024-09-05	

tail(pm_2005)

	State.Code	County.Code	Site.Num	Parameter.Code	POC	Latitude	Longitude
145908	CC	11	2	88101	1	49.13514	-102.9849
145909	CC	11	2	88101	1	49.13514	-102.9849
145910	CC	11	2	88101	1	49.13514	-102.9849
145911	CC	11	2	88101	1	49.13514	-102.9849
145912	CC	11	2	88101	1	49.13514	-102.9849
145913	CC	11	2	88101	1	49.13514	-102.9849
	Datum		Parameter.Name	Sample.Duration	Pollutant.Standard		
145908	WGS84	PM2.5 - Local	Conditions	24 HOUR	PM25	24-hour	2012
145909	WGS84	PM2.5 - Local	Conditions	24 HOUR	PM25	24-hour	2012
145910	WGS84	PM2.5 - Local	Conditions	24 HOUR	PM25	24-hour	2012
145911	WGS84	PM2.5 - Local	Conditions	24 HOUR	PM25	24-hour	2012
145912	WGS84	PM2.5 - Local	Conditions	24 HOUR	PM25	24-hour	2012
145913	WGS84	PM2.5 - Local	Conditions	24 HOUR	PM25	24-hour	2012
	Date.Local	Units.of.Measure		Event.Type	Observation.Count		
145908	2005-11-24	Micrograms/cubic meter (LC)		None	1		
145909	2005-11-30	Micrograms/cubic meter (LC)		None	1		
145910	2005-12-07	Micrograms/cubic meter (LC)		None	1		
145911	2005-12-12	Micrograms/cubic meter (LC)		None	1		
145912	2005-12-18	Micrograms/cubic meter (LC)		None	1		
145913	2005-12-30	Micrograms/cubic meter (LC)		None	1		
	Observation.Percent	Arithmetic.Mean	X1st.Max.Value	X1st.Max.Hour	AQI		
145908	100	3.8	3.8	0	21		
145909	100	4.0	4.0	0	22		

2/8/26, 5:12 PM

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145910	100	6.6	6.6	0	37
145911	100	4.0	4.0	0	22
145912	100	5.1	5.1	0	28
145913	100	9.1	9.1	15	51
Method.Code			Method.Name		
145908	143 R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric				
145909	143 R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric				
145910	143 R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric				
145911	143 R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric				
145912	143 R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric				
145913	143 R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric				
Local.Site.Name		Address	State.Name	County.Name	City.Name
145908	ESTEVAN, SK		Canada	Saskatchewan	Not in a city
145909	ESTEVAN, SK		Canada	Saskatchewan	Not in a city
145910	ESTEVAN, SK		Canada	Saskatchewan	Not in a city
145911	ESTEVAN, SK		Canada	Saskatchewan	Not in a city
145912	ESTEVAN, SK		Canada	Saskatchewan	Not in a city
145913	ESTEVAN, SK		Canada	Saskatchewan	Not in a city
CBSA.Name		Date.of.Last.Change			
145908	2024-05-19				
145909	2024-05-19				
145910	2024-05-19				
145911	2024-05-19				
145912	2024-05-19				
145913	2024-05-19				

head(pm_2025)

State.Code		County.Code	Site.Num	Parameter.Code	POC	Latitude	Longitude	Datum
1	1	3	10	88101	3	30.49748	-87.88026	NAD83
2	1	3	10	88101	3	30.49748	-87.88026	NAD83
3	1	3	10	88101	3	30.49748	-87.88026	NAD83
4	1	3	10	88101	3	30.49748	-87.88026	NAD83
5	1	3	10	88101	3	30.49748	-87.88026	NAD83
6	1	3	10	88101	3	30.49748	-87.88026	NAD83
Parameter.Name		Sample.Duration	Pollutant.Standard		Date.Local			
1	PM2.5 - Local Conditions	1 HOUR			2025-01-01			
2	PM2.5 - Local Conditions	1 HOUR			2025-01-02			
3	PM2.5 - Local Conditions	1 HOUR			2025-01-03			
4	PM2.5 - Local Conditions	1 HOUR			2025-01-04			
5	PM2.5 - Local Conditions	1 HOUR			2025-01-05			
6	PM2.5 - Local Conditions	1 HOUR			2025-01-06			
Units.of.Measure		Event.Type	Observation.Count	Observation.Percent				
1	Micrograms/cubic meter (LC)	None	24	100				
2	Micrograms/cubic meter (LC)	None	24	100				
3	Micrograms/cubic meter (LC)	None	24	100				
4	Micrograms/cubic meter (LC)	None	24	100				
5	Micrograms/cubic meter (LC)	None	24	100				
6	Micrograms/cubic meter (LC)	None	23	96				
Arithmetic.Mean X1st.Max.Value X1st.Max.Hour AQI Method.Code								

1	3.625000	19	21	NA	209
2	6.791667	26	23	NA	209
3	9.958333	25	1	NA	209
4	5.500000	11	19	NA	209
5	5.500000	12	14	NA	209
6	2.565217	16	2	NA	209

Method.Name					
1	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation				
2	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation				
3	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation				
4	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation				
5	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation				
6	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation				

Local.Site.Name				Address
1	FAIRHOPE, Alabama FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA			
2	FAIRHOPE, Alabama FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA			
3	FAIRHOPE, Alabama FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA			
4	FAIRHOPE, Alabama FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA			
5	FAIRHOPE, Alabama FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA			
6	FAIRHOPE, Alabama FAIRHOPE HIGH SCHOOL, 1 PIRATE DRIVE, FAIRHOPE, ALABAMA			

State.Name	County.Name	City.Name	CBSA.Name
1	Alabama	Baldwin	Fairhope Daphne-Fairhope-Foley, AL
2	Alabama	Baldwin	Fairhope Daphne-Fairhope-Foley, AL
3	Alabama	Baldwin	Fairhope Daphne-Fairhope-Foley, AL
4	Alabama	Baldwin	Fairhope Daphne-Fairhope-Foley, AL
5	Alabama	Baldwin	Fairhope Daphne-Fairhope-Foley, AL
6	Alabama	Baldwin	Fairhope Daphne-Fairhope-Foley, AL

Date.of.Last.Change	
1	2025-10-29
2	2025-10-29
3	2025-10-29
4	2025-10-29
5	2025-10-29
6	2025-10-29

tail(pm_2025)

State.Code	County.Code	Site.Num	Parameter.Code	POC	Latitude	Longitude
466755	80	26	6	88101	1 31.29129	-110.9515
466756	80	26	6	88101	1 31.29129	-110.9515
466757	80	26	6	88101	1 31.29129	-110.9515
466758	80	26	6	88101	1 31.29129	-110.9515
466759	80	26	6	88101	1 31.29129	-110.9515
466760	80	26	6	88101	1 31.29129	-110.9515

Datum	Parameter.Name	Sample.Duration	Pollutant.Standard
466755	WGS84 PM2.5 - Local Conditions	24-HR BLK AVG	PM25 24-hour 2012
466756	WGS84 PM2.5 - Local Conditions	24-HR BLK AVG	PM25 24-hour 2012
466757	WGS84 PM2.5 - Local Conditions	24-HR BLK AVG	PM25 24-hour 2012
466758	WGS84 PM2.5 - Local Conditions	24-HR BLK AVG	PM25 24-hour 2012
466759	WGS84 PM2.5 - Local Conditions	24-HR BLK AVG	PM25 24-hour 2012

466760	WGS84 PM2.5 - Local Conditions	24-HR BLK AVG	PM25	24-hour	2012
	Date.Local	Units.of.Measure	Event.Type	Observation.Count	
466755	2025-09-25	Micrograms/cubic meter (LC)	None	1	
466756	2025-09-26	Micrograms/cubic meter (LC)	None	1	
466757	2025-09-27	Micrograms/cubic meter (LC)	None	1	
466758	2025-09-28	Micrograms/cubic meter (LC)	None	1	
466759	2025-09-29	Micrograms/cubic meter (LC)	None	1	
466760	2025-09-30	Micrograms/cubic meter (LC)	None	1	
	Observation.Percent	Arithmetic.Mean	X1st.Max.Value	X1st.Max.Hour	AQI
466755	100	9.9	9.9	0	52
466756	100	3.5	3.5	0	19
466757	100	7.9	7.9	0	44
466758	100	11.2	11.2	0	55
466759	100	15.3	15.3	0	63
466760	100	17.6	17.6	0	67
	Method.Code				
466755	638				
466756	638				
466757	638				
466758	638				
466759	638				
466760	638				
				Method.Name	
466755	Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy				
466756	Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy				
466757	Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy				
466758	Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy				
466759	Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy				
466760	Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy				
	Local.Site.Name				
466755	Nogales Sonora Institute ITN				
466756	Nogales Sonora Institute ITN				
466757	Nogales Sonora Institute ITN				
466758	Nogales Sonora Institute ITN				
466759	Nogales Sonora Institute ITN				
466760	Nogales Sonora Institute ITN				
				Address	
466755	Avenida Instituto Tecnologico #911, Granja, 84065 Nogales, Son., Mexico				
466756	Avenida Instituto Tecnologico #911, Granja, 84065 Nogales, Son., Mexico				
466757	Avenida Instituto Tecnologico #911, Granja, 84065 Nogales, Son., Mexico				
466758	Avenida Instituto Tecnologico #911, Granja, 84065 Nogales, Son., Mexico				
466759	Avenida Instituto Tecnologico #911, Granja, 84065 Nogales, Son., Mexico				
466760	Avenida Instituto Tecnologico #911, Granja, 84065 Nogales, Son., Mexico				
	State.Name	County.Name	City.Name	CBSA.Name	
466755	Country Of Mexico	SONORA	Not in a city		
466756	Country Of Mexico	SONORA	Not in a city		
466757	Country Of Mexico	SONORA	Not in a city		
466758	Country Of Mexico	SONORA	Not in a city		
466759	Country Of Mexico	SONORA	Not in a city		
466760	Country Of Mexico	SONORA	Not in a city		
	Date.of.Last.Change				

466755	2025-10-22
466756	2025-10-22
466757	2025-10-22
466758	2025-10-22
466759	2025-10-22
466760	2025-10-22

EDA Step 5: Visualize the distribution of PM_{2.5}

```
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

```
library(ggplot2)
```

```
df <- pm_2005
```

```
df2 <- df %>%
  mutate(
    Date = as.Date(Date.Local),
    PM_2_5 = as.numeric(Arithmetic.Mean)
  )
```

```
# Michigan-only (state code 26)
```

```
mi <- df2 %>%
  filter(State.Code == 26)
```

```
# Numeric summary (all)
```

```
summary(df2$PM_2_5)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.00	6.80	10.70	12.73	16.40	132.60

```
quantile(df2$PM_2_5, probs = c(0, .01, .05, .10, .25, .50, .75, .90, .95, .99, 1), na.rm = TRUE)
```

0%	1%	5%	10%	25%	50%	75%	90%	95%	99%	100%
0.0	1.9	3.3	4.4	6.8	10.7	16.4	23.8	29.2	41.6	132.6

```
mean(df2$PM_2_5, na.rm = TRUE)
```

```
[1] 12.73335
```

```
sd(df2$PM_2_5, na.rm = TRUE)
```

```
[1] 8.487454
```

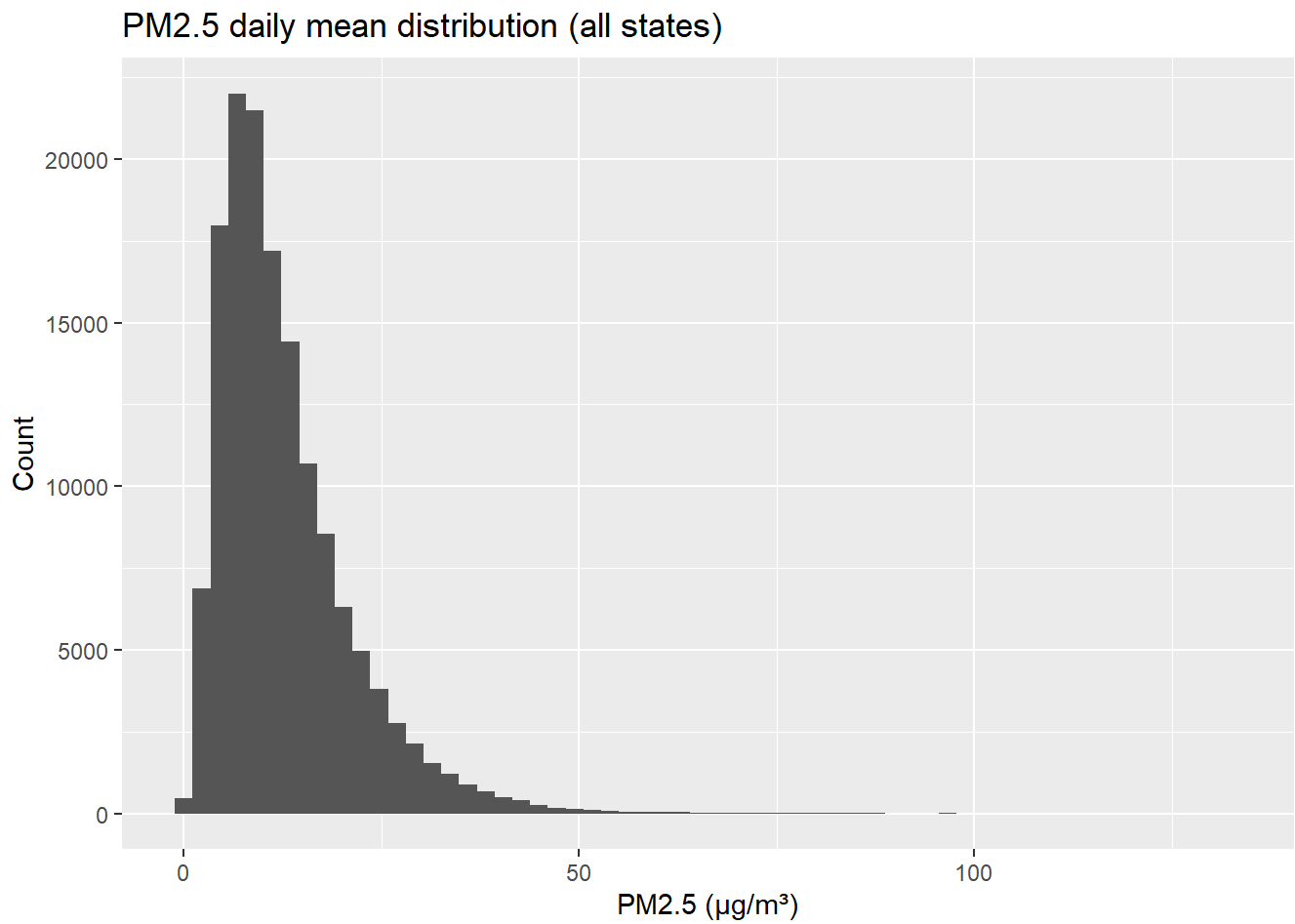
```
# Numeric summary (Michigan)  
summary(mi$PM_2_5)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.70	6.50	11.20	13.99	19.00	81.60

```
quantile(mi$PM_2_5, probs = c(0, .01, .05, .10, .25, .50, .75, .90, .95, .99, 1), na.rm = TRUE)
```

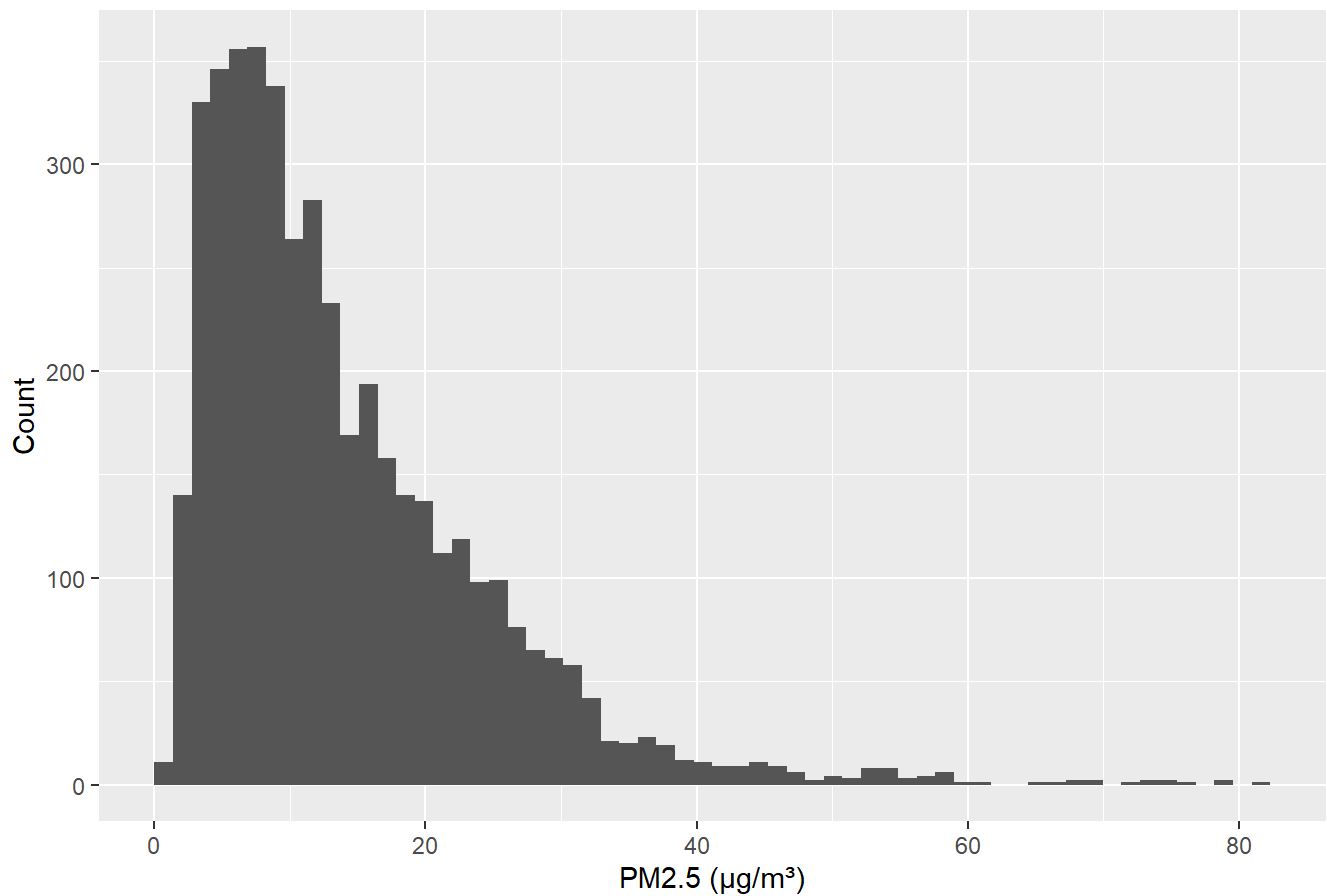
0%	1%	5%	10%	25%	50%	75%	90%	95%	99%	100%
0.70	2.00	3.20	4.00	6.50	11.20	19.00	27.20	32.35	52.30	81.60

```
ggplot(df2, aes(x = PM_2_5)) +  
  geom_histogram(bins = 60) +  
  labs(title = "PM2.5 daily mean distribution (all states)", x = "PM2.5 (µg/m³)", y = "Count")
```

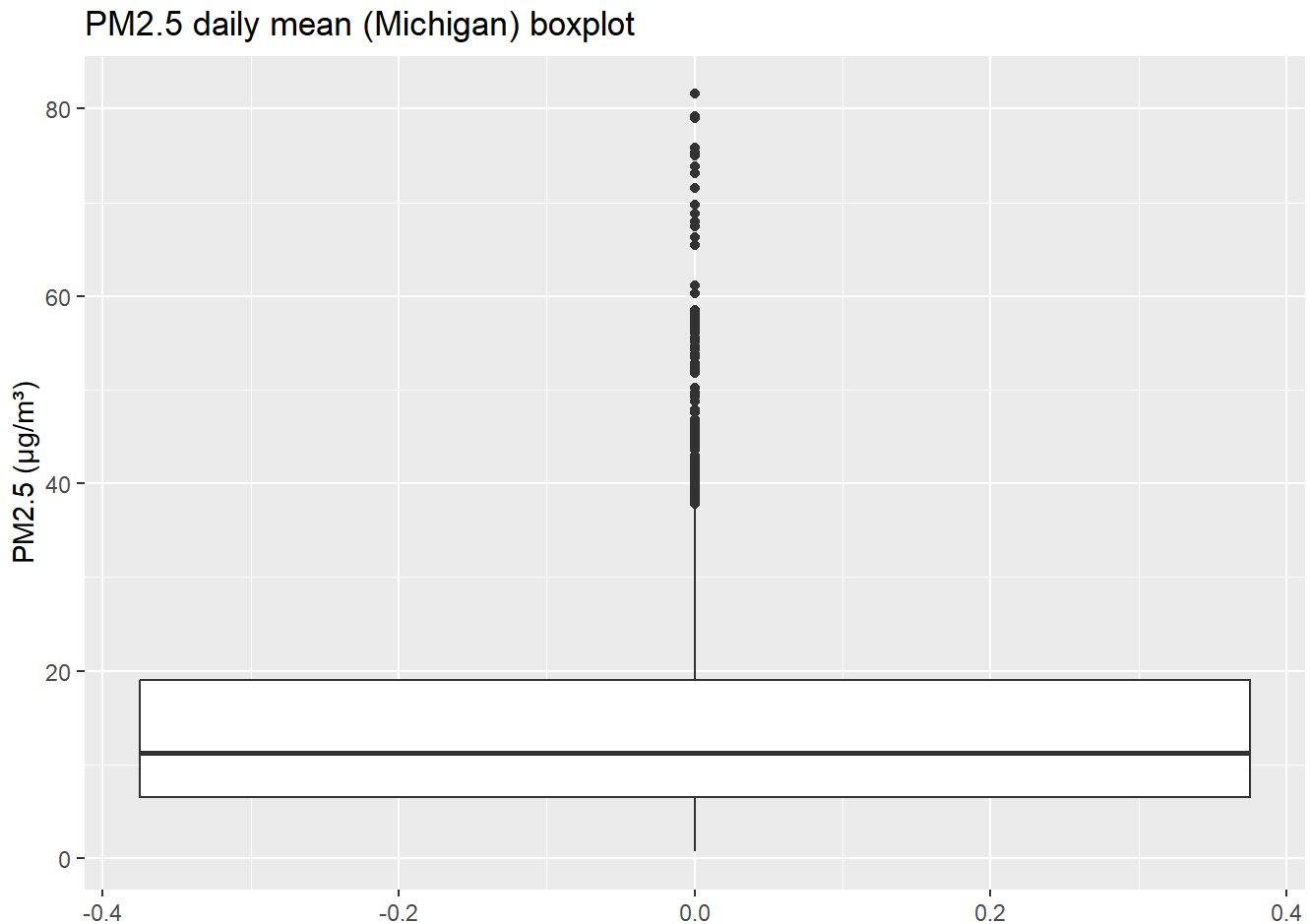


```
ggplot(mi, aes(x = PM_2_5)) +  
  geom_histogram(bins = 60) +  
  labs(title = "PM2.5 daily mean distribution (Michigan)", x = "PM2.5 (µg/m³)", y = "Count")
```

PM2.5 daily mean distribution (Michigan)



```
# Boxplot (Michigan) - good for outliers / skew
ggplot(mi, aes(y = PM_2_5)) +
  geom_boxplot() +
  labs(title = "PM2.5 daily mean (Michigan) boxplot", y = "PM2.5 (µg/m³)")
```



EDA Summary

After reading the data into R, we find that both the 2005 and 2025 PM2.5 data have 29 variables (columns), which have the same names between the datasets. The 2005 data have 145,913 records (rows), while the 2025 data have 466,760 records.

The variables in the data include important information such as state, county, latitude, longitude, date, units of measurement, and the most important variable: the PM2.5 measurement, recorded as "Arithmetic.Mean". Some variables encode the same element in different ways, such as "State.Code" and "State.Name", although they are both the "chr" data type in this case. Other variables like latitude, longitude, and Arithmetic.Mean have the "num" datatype to store floating-point numbers.

From examining the top and bottom of each dataset, many of the values are grouped into similar sets, so the values at the top and bottom are frequently the same. The exception is Arithmetic.Mean, which always sees significant changes between records.

From observing the histograms of the distribution of PM2.5, we find that in both Michigan and overall, the distribution is right-skewed, with the mean located at roughly $8 \mu\text{g}/\text{m}^3$. The frequency of PM2.5 concentrations decreases sharply after this point, with a very small number of records surpassing $50 \mu\text{g}/\text{m}^3$. The boxplot of Michigan records confirms that any records above $40 \mu\text{g}/\text{m}^3$ are considered outliers.

2. Combine data frames into one

```
library(dplyr)

clean_pm <- function(df) {
  df %>%
    mutate(
      Date = as.Date(Date.Local),
      year = as.integer(format(Date, "%Y")),
      pm25 = as.numeric(Arithmetic.Mean),
      state = State.Name,
      county = County.Name,
      site = Site.Num
    ) %>%
    select(state, county, site, year, Date, pm25,
           Latitude, Longitude, Method.Name, Units.of.Measure) %>%
    arrange(state, county, site, Date)
}

pm_all <- bind_rows(
  clean_pm(pm_2005),
  clean_pm(pm_2025)
)

# Michigan only
pm_mi <- pm_all %>% filter(state == 26)
```

```
dplyr::count(pm_all, year)
```

```
  year      n
1 2005 145913
2 2025 466760
```

```
summary(pm_all$pm25)
```

```
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
-5.733   4.500   6.900   8.656  10.662  288.500
```

```
str(pm_all)
```

```
'data.frame':  612673 obs. of  10 variables:
 $ state      : chr  "Alabama" "Alabama" "Alabama" "Alabama" ...
 $ county     : chr  "Baldwin" "Baldwin" "Baldwin" "Baldwin" ...
 $ site       : int   10 10 10 10 10 10 10 10 10 10 ...
 $ year       : int   2005 2005 2005 2005 2005 2005 2005 2005 2005 2005 ...
 $ Date       : Date, format: "2005-01-01" "2005-01-04" ...
 $ pm25       : num   9 7.7 7.3 8.7 3.3 7.7 16.6 7.9 15.1 23.4 ...
 $ Latitude   : num  30.5 30.5 30.5 30.5 30.5 ...
```

```
$ Longitude      : num  -87.9 -87.9 -87.9 -87.9 -87.9 ...
$ Method.Name    : chr   "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" "R & P Model
2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" "R & P Model 2025 PM2.5 Sequential w/WINS -
GRAVIMETRIC" "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC" ...
$ Units.of.Measure: chr   "Micrograms/cubic meter (LC)" "Micrograms/cubic meter (LC)"
"Micrograms/cubic meter (LC)" "Micrograms/cubic meter (LC)" ...
```

```
head(pm_all)
```

	state	county	site	year	Date	pm25	Latitude	Longitude
1	Alabama	Baldwin	10	2005	2005-01-01	9.0	30.49748	-87.88026
2	Alabama	Baldwin	10	2005	2005-01-04	7.7	30.49748	-87.88026
3	Alabama	Baldwin	10	2005	2005-01-07	7.3	30.49748	-87.88026
4	Alabama	Baldwin	10	2005	2005-01-10	8.7	30.49748	-87.88026
5	Alabama	Baldwin	10	2005	2005-01-13	3.3	30.49748	-87.88026
6	Alabama	Baldwin	10	2005	2005-01-16	7.7	30.49748	-87.88026

	Method.Name
1	R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC
2	R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC
3	R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC
4	R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC
5	R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC
6	R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC

	Units.of.Measure
1	Micrograms/cubic meter (LC)
2	Micrograms/cubic meter (LC)
3	Micrograms/cubic meter (LC)
4	Micrograms/cubic meter (LC)
5	Micrograms/cubic meter (LC)
6	Micrograms/cubic meter (LC)

```
tail(pm_all)
```

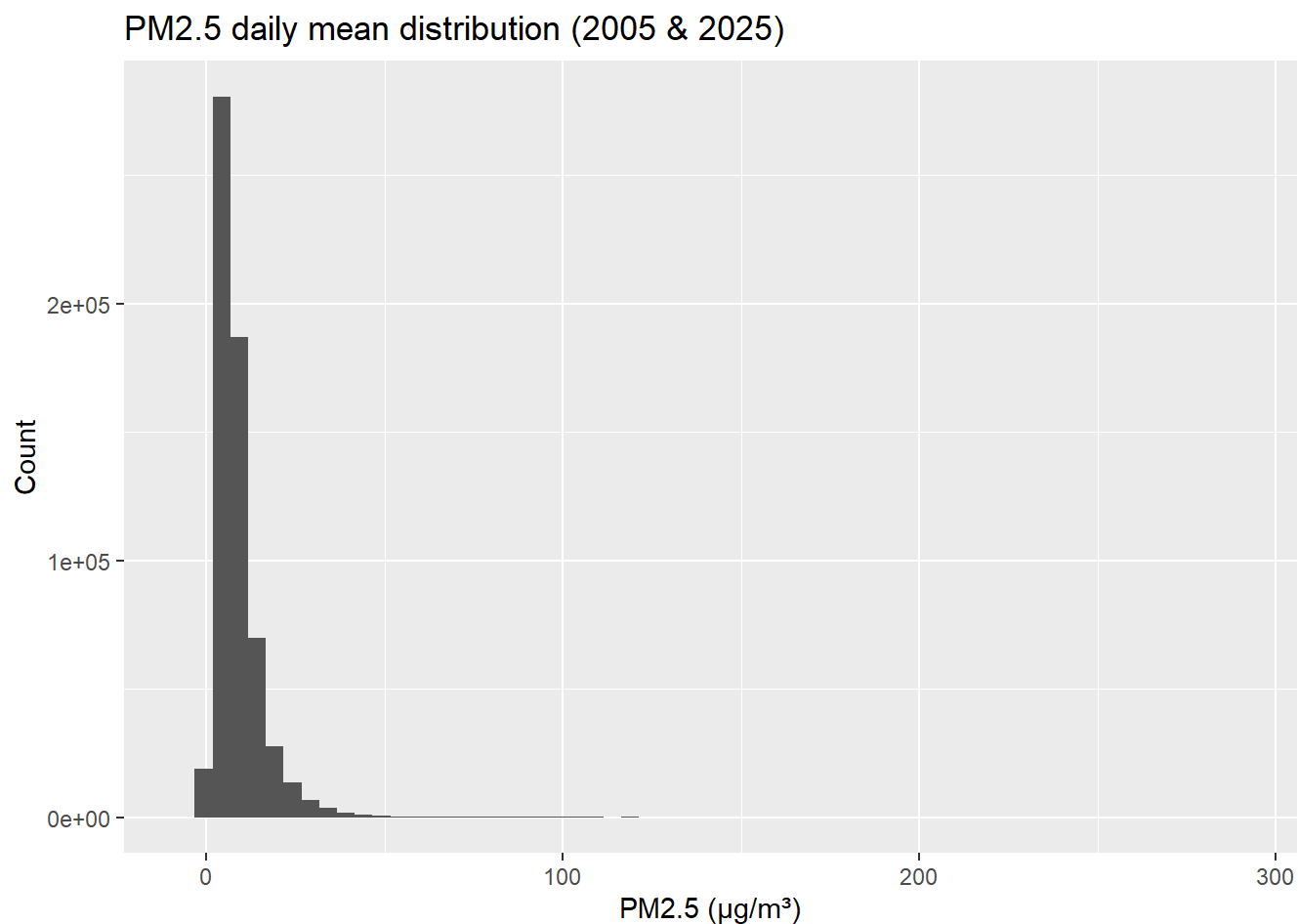
	state	county	site	year	Date	pm25	Latitude	Longitude
612668	Wyoming	Washakie	2	2025	2025-07-29	4.141667	44.00792	-107.957
612669	Wyoming	Washakie	2	2025	2025-07-29	4.100000	44.00792	-107.957
612670	Wyoming	Washakie	2	2025	2025-07-30	4.687500	44.00792	-107.957
612671	Wyoming	Washakie	2	2025	2025-07-30	4.600000	44.00792	-107.957
612672	Wyoming	Washakie	2	2025	2025-07-31	4.825000	44.00792	-107.957
612673	Wyoming	Washakie	2	2025	2025-07-31	4.800000	44.00792	-107.957

	Method.Name
612668	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation
612669	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation
612670	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation
612671	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation
612672	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation
612673	Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation

	Units.of.Measure
612668	Micrograms/cubic meter (LC)

612669 Micrograms/cubic meter (LC)
612670 Micrograms/cubic meter (LC)
612671 Micrograms/cubic meter (LC)
612672 Micrograms/cubic meter (LC)
612673 Micrograms/cubic meter (LC)

```
ggplot(pm_all, aes(x = pm25)) +  
  geom_histogram(bins = 60) +  
  labs(title = "PM2.5 daily mean distribution (2005 & 2025)", x = "PM2.5 (µg/m³)", y = "Count")
```



3. Create leaflet map to show monitoring site locations

```
library(dplyr)  
library(leaflet)  
  
# Filter to Michigan data  
df <- pm_all %>% filter(state == "Michigan")  
  
# One row per site per year  
sites_year <- df %>%  
  filter(!is.na(Latitude), !is.na(Longitude)) %>%  
  distinct(year, state, county, site, Latitude, Longitude)
```

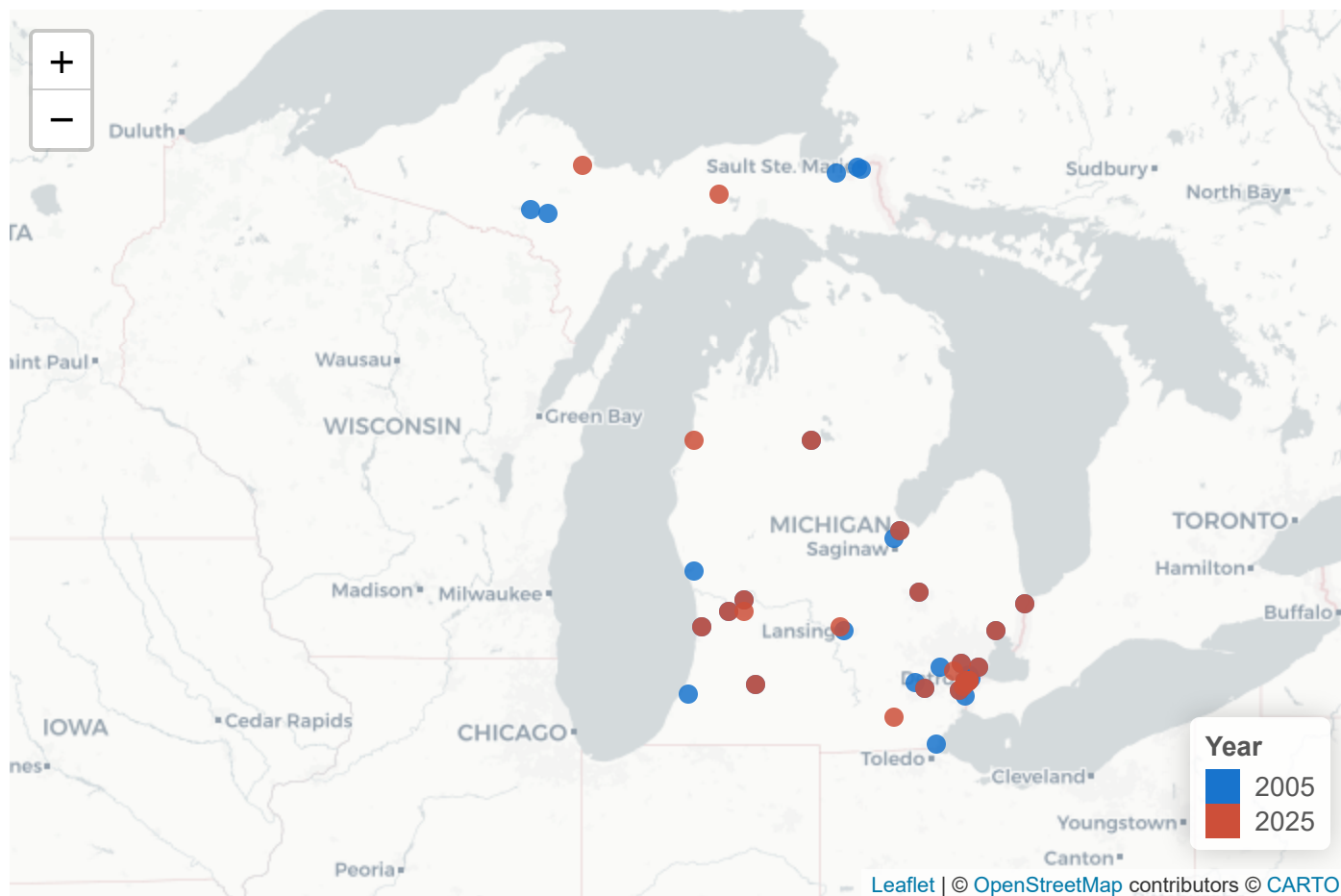


```

pal <- colorFactor(palette = c("dodgerblue3", "tomato3"), domain = sort(unique(sites_year$year)))

m <- leaflet(sites_year) %>%
  addProviderTiles(providers$CartoDB.Positron) %>%
  addCircleMarkers(
    lng = ~Longitude, lat = ~Latitude,
    radius = 5, stroke = FALSE, fillOpacity = 0.85,
    color = ~pal(year),
    popup = ~paste0(
      "<b>Year:</b> ", year, "<br/>",
      "<b>Site:</b> ", state, "-", county, "-", site, "<br/>",
      "<b>Lat/Lon:</b> ", round(Latitude, 4), ", ", round(Longitude, 4)
    )
  ) %>%
  addLegend(
    position = "bottomright",
    pal = pal, values = ~year,
    title = "Year", opacity = 1
  )
m

```



```

# Counts by year
sites_year %>%

```

```
count(year, name = "n_sites")
```

```
year n_sites
1 2005      31
2 2025      26
```

```
# Bounding box by year (rough geographic footprint)
sites_year %>%
  group_by(year) %>%
  summarize(
    n_sites = n(),
    lat_min = min(Latitude), lat_max = max(Latitude),
    lon_min = min(Longitude), lon_max = max(Longitude),
    .groups = "drop"
  )
```

```
# A tibble: 2 × 6
```

```
year n_sites lat_min lat_max lon_min lon_max
<int> <int> <dbl> <dbl> <dbl> <dbl>
1 2005      31  41.8  46.5 -88.1 -82.5
2 2025      26  42.0  46.5 -87.5 -82.5
```

```
# Sites present in both years vs. only one year
site_id <- sites_year %>%
  mutate(site_id = paste(state, county, site, sep = "-")) %>%
  distinct(year, site_id)

site_id %>%
  count(site_id) %>%
  count(n, name = "n_sites") %>%
  mutate(present_in_years = n) %>%
  arrange(desc(present_in_years))
```

```
n n_sites present_in_years
1 2      15              2
2 1      27              1
```

From the map, we find the the monitoring sites are not evenly spread across Michigan. Instead, they have the highest concentration in Southeeast Michigan (the Detroit - Ann Arbor - Downriver area) for both 2005 and 2025. This makes sense, as this would be the area in Michigan where the population - and thus the emission source - is the highest. There are additional clusters of sites in central/east-central Michigan, such as around Lansing and the Saginaw / Bay City / Flint region. Meanwhile, there are very few sites in western and northern Michigan (which were only introduced as of 2025), and a small amount in the Upper Peninsula. Overall, while the number of monitoring sites increased from 2005 to 2025, the spatial distribution of the sites across Michigan remained mostly unchanged.

4. Check for data issues

Identify missing or implausible values

```
library(dplyr)
```

```
df <- pm_all %>%
  mutate(
    Date = as.Date(Date),
    year = as.integer(year),
    pm25 = as.numeric(pm25)
  )
```

```
# 1) Missingness
df %>%
  summarize(
    n = n(),
    pm25_missing = sum(is.na(pm25)),
    pm25_missing_pct = mean(is.na(pm25)) * 100
  )
```

	n	pm25_missing	pm25_missing_pct
1	612673	0	0

```
df %>%
  group_by(year) %>%
  summarize(
    n = n(),
    pm25_missing = sum(is.na(pm25)),
    pm25_missing_pct = mean(is.na(pm25)) * 100,
    .groups = "drop"
  )
```

```
# A tibble: 2 × 4
  year      n pm25_missing pm25_missing_pct
<int> <int>      <int>          <dbl>
1  2005 145913          0              0
2  2025 466760          0              0
```

We find that there are no values in the dataset that are completely missing.

```
# 2) Implausible values (negative, extremely large)
df %>%
  summarize(
    n_neg = sum(pm25 < 0, na.rm = TRUE),
    n_zero = sum(pm25 == 0, na.rm = TRUE),
    n_gt_500 = sum(pm25 > 500, na.rm = TRUE),
    n_gt_1000 = sum(pm25 > 1000, na.rm = TRUE)
  )
```

```
n_neg n_zero n_gt_500 n_gt_1000
1 1734 338 0 0
```

```
# Table of implausible PM25 measurements
df %>%
  filter(pm25 < 0 | pm25 > 500) %>%
  select(state, county, site, year, Date, pm25) %>%
  arrange(desc(pm25)) %>%
  head(50)
```

	state	county	site	year	Date	pm25
1	Wyoming	Washakie	2	2025	2025-03-08	-0.004167
2	California	Kern	20	2025	2025-02-07	-0.008333
3	Texas	Potter	1025	2025	2025-01-02	-0.008333
4	California	Riverside	1016	2025	2025-01-02	-0.008696
5	New Jersey	Atlantic	6	2025	2025-01-09	-0.011765
6	Oregon	Marion	41	2025	2025-04-28	-0.012500
7	New Mexico	Taos	5	2025	2025-03-15	-0.020833
8	Oregon	Lane	2013	2025	2025-04-07	-0.020833
9	Oregon	Lane	2013	2025	2025-05-17	-0.020833
10	California	Imperial	7	2025	2025-01-27	-0.025000
11	Montana	Lewis and Clark	4	2025	2025-03-23	-0.025000
12	Oregon	Lane	60	2025	2025-03-21	-0.025000
13	Texas	Potter	1025	2025	2025-03-30	-0.025000
14	Nebraska	Gage	5	2025	2025-06-19	-0.029167
15	South Dakota	Minnehaha	9	2025	2025-02-28	-0.029167
16	California	Riverside	1016	2025	2025-02-15	-0.033333
17	Oregon	Deschutes	120	2025	2025-04-22	-0.033333
18	Oregon	Lane	60	2025	2025-01-03	-0.033333
19	Texas	Potter	1025	2025	2025-02-08	-0.033333
20	California	Nevada	1001	2025	2025-01-02	-0.034783
21	Montana	Lewis and Clark	4	2025	2025-05-20	-0.037500
22	Utah	Grand	7	2025	2025-03-14	-0.037500
23	Wyoming	Teton	1006	2025	2025-05-06	-0.037500
24	California	Colusa	1002	2025	2025-01-13	-0.039130
25	Alaska	Matanuska-Susitna	10	2025	2025-01-15	-0.041667
26	Alaska	Matanuska-Susitna	10	2025	2025-05-15	-0.041667
27	California	Colusa	7	2025	2025-02-17	-0.041667
28	California	Colusa	7	2025	2025-02-18	-0.041667
29	California	Los Angeles	4009	2025	2025-02-13	-0.041667
30	California	San Luis Obispo	8002	2025	2025-07-28	-0.041667
31	California	Ventura	9	2025	2025-03-31	-0.041667
32	California	Ventura	2002	2025	2025-01-23	-0.041667
33	Hawaii	Hawaii	5	2025	2025-06-17	-0.041667
34	Maine	Cumberland	29	2025	2025-05-19	-0.041667
35	Montana	Beaverhead	3	2025	2025-01-22	-0.041667
36	North Carolina	Durham	15	2025	2025-08-22	-0.041667
37	Oregon	Lane	1013	2025	2025-03-02	-0.041667
38	Washington	Stevens	5	2025	2025-06-21	-0.041667
39	Washington	Stevens	5	2025	2025-06-30	-0.041667

40	Wyoming	Albany	12	2025	2025-01-05	-0.041667
41	Wyoming	Albany	12	2025	2025-02-15	-0.041667
42	Wyoming	Fremont	99	2025	2025-02-17	-0.041667
43	Wyoming	Fremont	232	2025	2025-02-07	-0.041667
44	Wyoming	Fremont	232	2025	2025-03-04	-0.041667
45	Wyoming	Park	1	2025	2025-01-26	-0.041667
46	Wyoming	Sweetwater	7	2025	2025-02-13	-0.041667
47	Montana	Missoula	24	2025	2025-05-13	-0.042857
48	California	Fresno	500	2025	2025-10-14	-0.043478
49	California	Los Angeles	9035	2025	2025-03-11	-0.043478
50	California	San Luis Obispo	2004	2025	2025-07-25	-0.043478

We find that some of the PM2.5 measurements are negative, and a small number are exactly 0. Since these are implausible and irrelevant, respectively, we filter these records out of the dataset:

```
df <- df %>% filter(pm25 > 0)

# Table of plausible PM25 measurements
df %>%
  select(state, county, site, year, Date, pm25) %>%
  arrange(desc(pm25)) %>%
  head(50)
```

	state	county	site	year	Date	pm25
1	Texas	El Paso	44	2025	2025-04-01	288.5000
2	Texas	El Paso	55	2025	2025-03-18	285.8235
3	Texas	El Paso	55	2025	2025-03-18	285.8000
4	New Mexico	Dona Ana	21	2025	2025-03-18	259.3913
5	New Mexico	Dona Ana	21	2025	2025-03-18	259.3000
6	Arizona	Maricopa	19	2025	2025-01-01	258.1917
7	Arizona	Maricopa	19	2025	2025-01-01	258.1000
8	New Mexico	Dona Ana	21	2025	2025-03-06	246.2792
9	New Mexico	Dona Ana	21	2025	2025-03-06	246.2000
10	New Mexico	Dona Ana	21	2025	2025-03-06	242.4208
11	New Mexico	Dona Ana	21	2025	2025-03-06	242.4000
12	Texas	El Paso	55	2025	2025-03-03	238.6667
13	Texas	El Paso	55	2025	2025-03-03	238.6000
14	New Mexico	Dona Ana	22	2025	2025-03-06	233.4542
15	New Mexico	Dona Ana	22	2025	2025-03-06	233.4000
16	Texas	El Paso	44	2025	2025-03-18	228.6500
17	Texas	El Paso	44	2025	2025-03-18	228.6000
18	New Mexico	Dona Ana	21	2025	2025-03-18	226.6870
19	New Mexico	Dona Ana	21	2025	2025-03-18	226.6000
20	New Mexico	Dona Ana	22	2025	2025-03-18	221.0250
21	New Mexico	Dona Ana	22	2025	2025-03-18	221.0000
22	Arizona	Maricopa	4003	2025	2025-01-01	210.3292
23	Arizona	Maricopa	4003	2025	2025-01-01	210.3000
24	North Dakota	Burke	4	2025	2025-05-31	210.3000
25	North Dakota	Burke	4	2025	2025-05-31	210.3000
26	North Dakota	Ward	3	2025	2025-05-31	208.4292
27	North Dakota	Ward	3	2025	2025-05-31	208.4000

28	North Dakota	Mercer	4	2025	2025-05-31	208.2708
29	North Dakota	Mercer	4	2025	2025-05-31	208.2000
30	North Dakota	Mercer	4	2025	2025-05-31	207.6500
31	North Dakota	Mercer	4	2025	2025-05-31	207.6000
32	New Mexico	Dona Ana	21	2025	2025-04-01	193.2083
33	New Mexico	Dona Ana	21	2025	2025-04-01	193.2000
34	New Mexico	Dona Ana	21	2025	2025-04-01	180.8083
35	New Mexico	Dona Ana	21	2025	2025-04-01	180.8000
36	Arizona	Maricopa	9812	2025	2025-01-01	167.4375
37	Arizona	Maricopa	9812	2025	2025-01-01	167.4000
38	Texas	El Paso	44	2025	2025-04-01	167.0091
39	Texas	El Paso	44	2025	2025-04-01	167.0000
40	North Dakota	Oliver	2	2025	2025-05-31	166.7708
41	North Dakota	Oliver	2	2025	2025-05-31	166.7000
42	Texas	El Paso	55	2025	2025-04-27	164.3750
43	Texas	El Paso	55	2025	2025-04-27	164.3000
44	Texas	El Paso	55	2025	2025-04-01	163.1364
45	Texas	El Paso	55	2025	2025-04-01	163.1000
46	New Mexico	Dona Ana	16	2025	2025-03-06	162.9875
47	New Mexico	Dona Ana	16	2025	2025-03-06	162.9000
48	New Mexico	Dona Ana	21	2025	2025-04-27	162.2792
49	New Mexico	Dona Ana	21	2025	2025-04-27	162.2000
50	New Mexico	Dona Ana	22	2025	2025-04-01	153.2875

```
# 3) Distribution checks by year
```

```
df %>%
```

```
  group_by(year) %>%
```

```
  summarize(
```

```
    n = sum(!is.na(pm25)),
```

```
    min = min(pm25, na.rm = TRUE),
```

```
    p1 = quantile(pm25, 0.01, na.rm = TRUE),
```

```
    p5 = quantile(pm25, 0.05, na.rm = TRUE),
```

```
    median = median(pm25, na.rm = TRUE),
```

```
    p95 = quantile(pm25, 0.95, na.rm = TRUE),
```

```
    p99 = quantile(pm25, 0.99, na.rm = TRUE),
```

```
    max = max(pm25, na.rm = TRUE),
```

```
    mean = mean(pm25, na.rm = TRUE),
```

```
    sd = sd(pm25, na.rm = TRUE),
```

```
    .groups = "drop"
```

```
)
```

```
# A tibble: 2 × 11
```

	year	n	min	p1	p5	median	p95	p99	max	mean	sd
	<int>	<int>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	2005	145856	0.1	1.9	3.3	10.7	29.2	41.6	133.	12.7	8.49
2	2025	464745	0.00417	0.989	2.1	6.19	16.2	27.1	288.	7.42	6.00

From this table, we see an overall level shift since the mean and median significantly decreased from 2005 to 2025; the mean decreased from 12.74 to 7.42, and the median decreased from 10.70 to 6.19. The highest percentiles are also lower in 2025 compared to 2005, which suggests that the overall PM2.5 concentrations

in the air were less in 2025. There are also outliers of very high maximum concentrations in both years (132.6 and 288.5 respectively), which could be caused by an extreme event such as a wildfire.

```
# 4) Missing days per site-year
site_days <- df %>%
  filter(!is.na(pm25)) %>%
  group_by(state, county, site, year) %>%
  summarize(
    n_days = n_distinct(Date),
    .groups = "drop"
  )

site_days %>%
  group_by(year) %>%
  summarize(
    n_site_years = n(),
    min_days = min(n_days),
    p25_days = quantile(n_days, 0.25),
    median_days = median(n_days),
    p75_days = quantile(n_days, 0.75),
    max_days = max(n_days),
    .groups = "drop"
  )
```

A tibble: 2 × 7

	year	n_site_years	min_days	p25_days	median_days	p75_days	max_days
	<int>	<int>	<int>	<dbl>	<int>	<dbl>	<int>
1	2005	1067	3	90	114	120	364
2	2025	1015	1	175	212	271	316

This shows that 2025 has roughly the same number of site-year pairs as 2005, with more frequent coverage per day in 2025 (higher median and percentiles), but some of the calendar year is not covered in 2025 (the maximum day is only 316 in 2025 compared to 364 in 2005). This makes sense because the 2025 data is only recorded as of November 24, 2025 according to the EPA website.

Check methods used for data collection

```
unique(pm_all$"Method.Name")
```

```
[1] "R & P Model 2025 PM2.5 Sequential w/WINS - GRAVIMETRIC"
[2] "BGI Models PQ200-VSCC or PQ200A-VSCC - Gravimetric"
[3] "BGI Model PQ200 PM2.5 Sampler w/WINS - GRAVIMETRIC"
[4] "Andersen RAAS2.5-300 PM2.5 SEQ w/WINS - GRAVIMETRIC"
[5] "R & P Model 2000 PM2.5 Sampler w/WINS - GRAVIMETRIC"
[6] "R & P Model 2000 PM-2.5 Air Sampler w/VSCC - Gravimetric"
[7] "R & P Model 2025 PM-2.5 Sequential Air Sampler w/VSCC - Gravimetric"
[8] "Andersen RAAS2.5-100 PM2.5 SAM w/WINS - GRAVIMETRIC"
[9] "Thermo Electron Model RAAS2.5-300 Sequential w/VSCC - Gravimetric"
```

```

[10] "Met One BAM-1022 Mass Monitor w/ VSCC or TE-PM2.5C - Beta Attenuation"
[11] "Teledyne T640X at 16.67 LPM w/Network Data Alignment enabled - Broadband spectroscopy"
[12] "Met One BAM-1020 Mass Monitor w/VSCC - Beta Attenuation"
[13] "Thermo Scientific TEOM 1405-DF Dichotomous FDMS - FDMS Gravimetric"
[14] "Met One E-FRM PM2.5 with VSCC - Gravimetric"
[15] "Teledyne T640X at 16.67 LPM - Broadband spectroscopy"
[16] "Teledyne T640X at 16.67 LPM (Corrected) - Broadband spectroscopy"
[17] "Thermo Scientific TEOM 1400 FDMS or 1405 8500C FDMS w/VSCC - FDMS Gravimetric"
[18] "Teledyne T640 at 5.0 LPM - Broadband spectroscopy"
[19] "Teledyne T640 at 5.0 LPM w/Network Data Alignment enabled - Broadband spectroscopy"
[20] "Met One E-SEQ-FRM PM2.5 with VSCC - Gravimetric"
[21] "GRIMM EDM Model 180 with naphion dryer - Laser Light Scattering"
[22] "Thermo Scientific 5014i or FH62C14-DHS w/VSCC - Beta Attenuation"
[23] "Thermo Scientific 1405-F FDMS w/VSCC - FDMS Gravimetric"
[24] "Tisch Model TE-Wilbur2.5 Low-Volume Sampler - Gravimetric"
[25] "Thermo Scientific Model 5030 SHARP w/VSCC - Beta Attenuation"

```

```

library(dplyr)

df <- pm_all %>%
  mutate(
    year = as.integer(year),
    method_name = Method.Name
  )

# 1) Proportion of missing Method.Name by year
miss_by_year <- df %>%
  group_by(year) %>%
  summarize(
    n = n(),
    n_missing = sum(is.na(method_name)),
    prop_missing = mean(is.na(method_name)),
    .groups = "drop"
  )

miss_by_year

```

```

# A tibble: 2 × 4
  year      n n_missing prop_missing
<int> <int>   <int>         <dbl>
1  2005 145913     0           0
2  2025 466760     0           0

```

```

library(stringr)

# 2) Distribution of Method.Name within each year (proportions among non-missing)
method_dist <- df %>%
  filter(!is.na(method_name)) %>%
  count(year, method_name, name = "n") %>%
  group_by(year) %>%

```



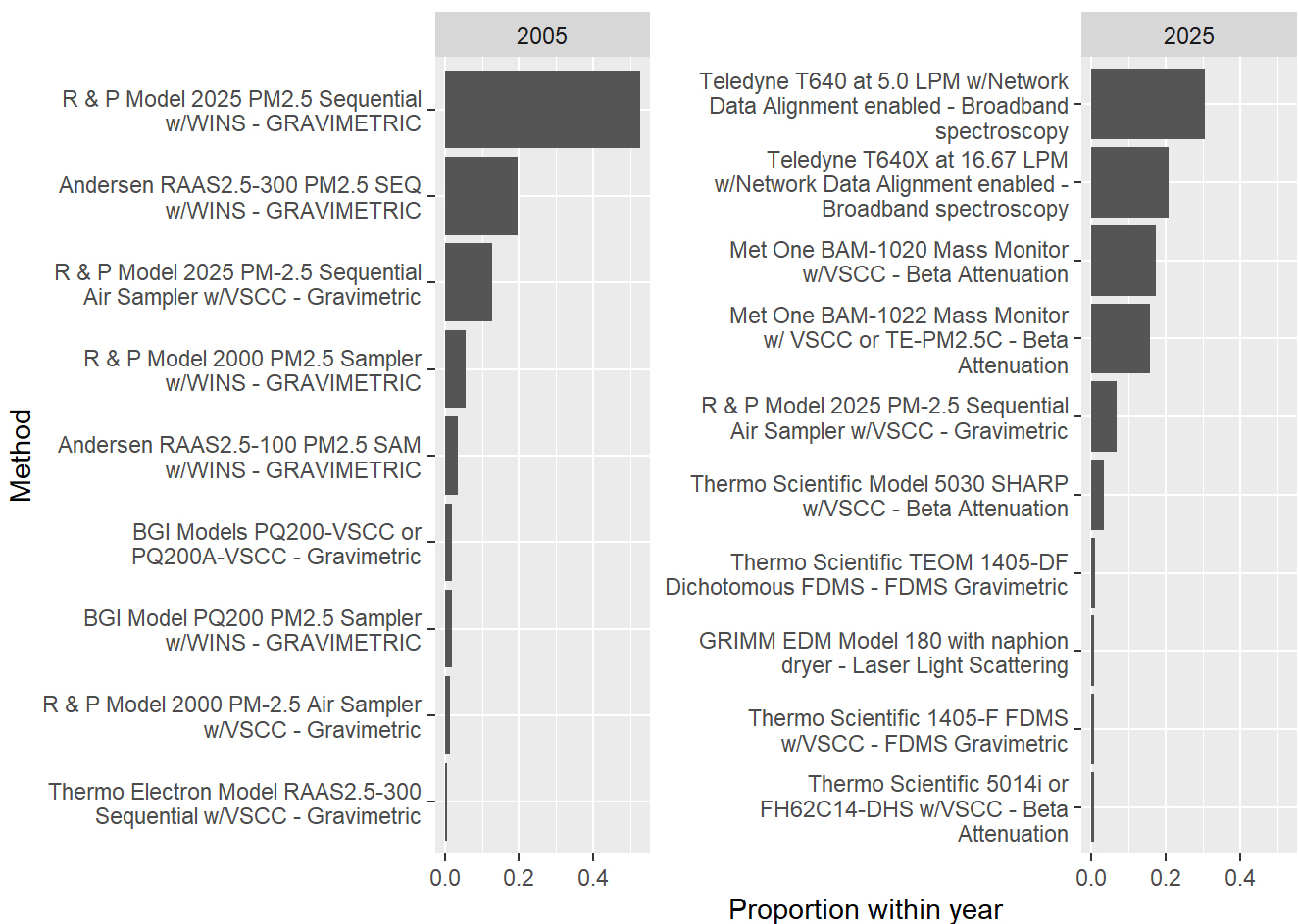
```

mutate(prop = n / sum(n)) %>%
arrange(year, desc(n)) %>%
ungroup()

# 3) Top 10 method codes per year
top10 <- method_dist %>%
  group_by(year) %>%
  slice_max(n, n = 10, with_ties = FALSE) %>%
  ungroup() %>%
  mutate(method_wrap = str_wrap(method_name, width = 35))

ggplot(top10, aes(x = prop, y = reorder(method_wrap, prop))) +
  geom_col() +
  facet_wrap(~ year, scales = "free_y") +
  labs(x = "Proportion within year", y = "Method")

```



From the [EPA PM2.5 Codetable](#), we find that the most common method names in 2005 correspond to codes 117 and 120, respectively, while the most common codes in 2025 are 636 and 638, respectively. However, the more helpful information is the proportion with respect to method names.

From 2005 to 2025, there is a big shift in the dominant measurement technology - instead of the overwhelming presence of gravimetric, filter-based samplers, there are more continuous and automated instruments (such as Teledyne for broadband spectroscopy and Met One for beta attenuation). Overall,

there is a more diverse set of instruments being used in 2025 compared to in 2005, with the most frequently used method in 2025 being a smaller proportion (about 30% of the time) compared to in 2005 (over 50% of the time). These temporal patterns are important to analyze because they reflect how methodologies have been modernized over the past 20 years to include new types of instruments. In this case, the new instruments can measure PM2.5 concentrations in real time instead of only collecting samples at scheduled times.

5. Visualize and summarize daily concentrations of PM2.5

```
# Standardize, restrict to Michigan, and keep only years of interest
library(dplyr)
library(ggplot2)
library(stringr)

pm_mi <- pm_all %>%
  mutate(
    state = if ("state" %in% names(.)) state else State.Code,
    county = if ("county" %in% names(.)) county else County.Code,
    site = if ("site" %in% names(.)) site else Site.Num,
    pm25 = if ("pm25" %in% names(.)) pm25 else as.numeric(Arithmetic.Mean),
    Date = if ("Date" %in% names(.)) as.Date(Date) else as.Date(Date.Local),
    year = if ("year" %in% names(.)) year else as.integer(format(Date, "%Y")),
    county_name = if ("County.Name" %in% names(.)) County.Name else NA_character_,
    city_name = if ("City.Name" %in% names(.)) City.Name else NA_character_,
    lat = as.numeric(if ("Latitude" %in% names(.)) Latitude else NA),
    lon = as.numeric(if ("Longitude" %in% names(.)) Longitude else NA)
  ) %>%
  filter(state == "Michigan", year %in% c(2005, 2025))
```

Level 1: PM2 concentrations in Michigan overall

```
state_summary <- pm_mi %>%
  group_by(year) %>%
  summarize(
    n_days = sum(!is.na(pm25)),
    mean = mean(pm25, na.rm = TRUE),
    median = median(pm25, na.rm = TRUE),
    p25 = quantile(pm25, 0.25, na.rm = TRUE),
    p75 = quantile(pm25, 0.75, na.rm = TRUE),
    p95 = quantile(pm25, 0.95, na.rm = TRUE),
    max = max(pm25, na.rm = TRUE),
    .groups = "drop"
  )

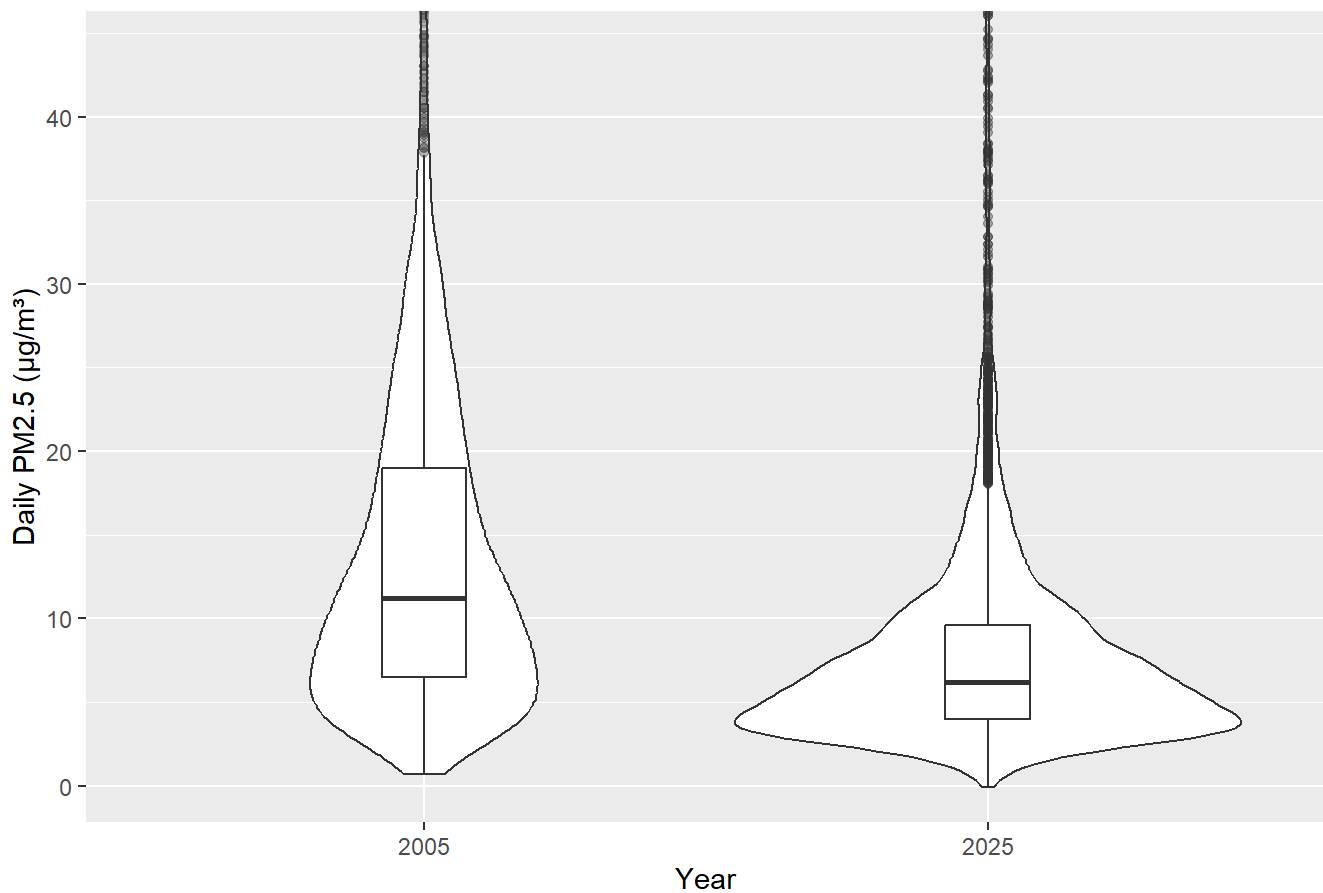
state_summary
```

A tibble: 2 × 8

	year	n_days	mean	median	p25	p75	p95	max
	<int>	<int>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	2005	4391	14.0	11.2	6.5	19	32.3	81.6
2	2025	10382	8.03	6.2	4	9.6	19.9	101.

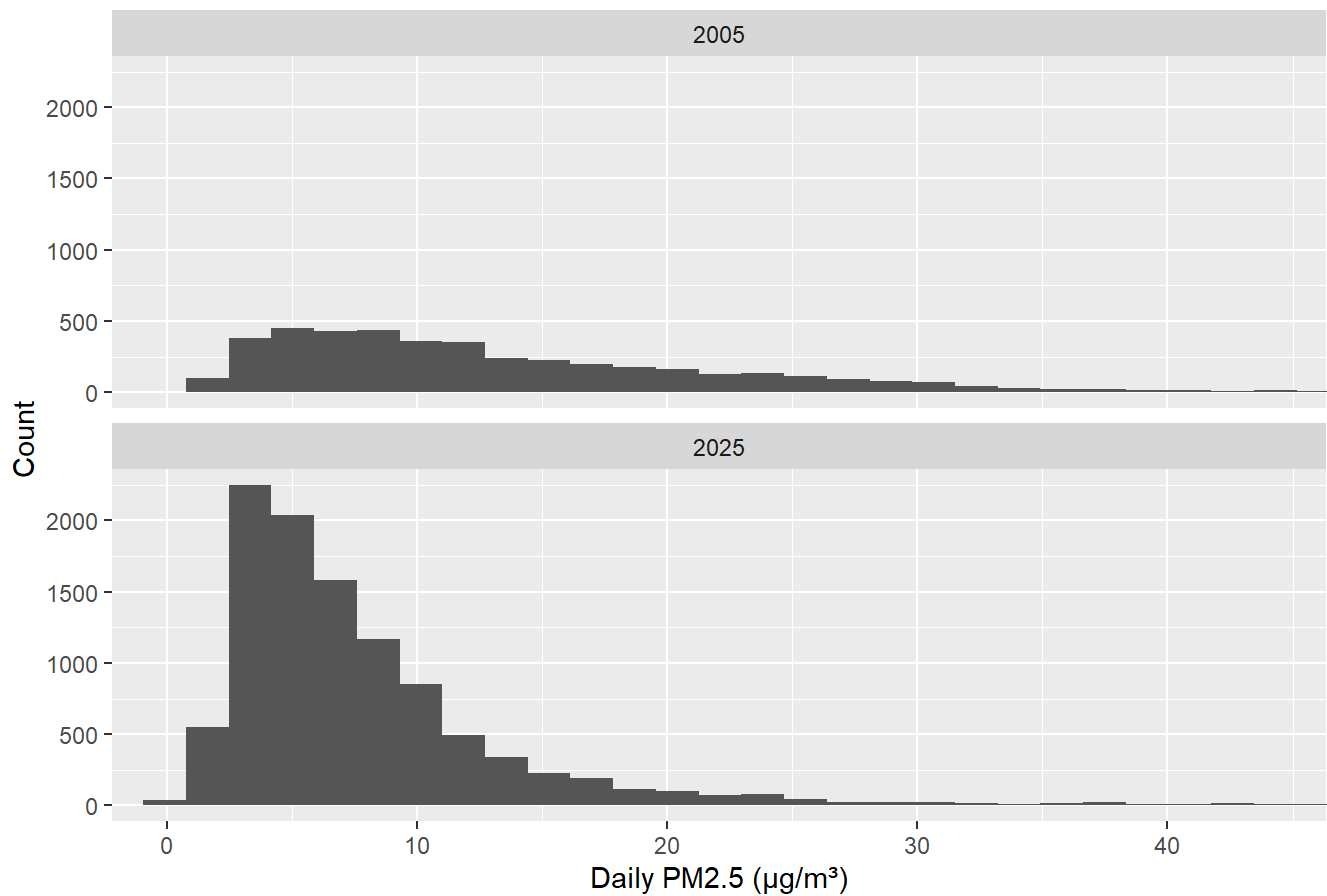
```
ggplot(pm_mi, aes(x = factor(year), y = pm25)) +
  geom_violin(trim = TRUE) +
  geom_boxplot(width = 0.15, outlier.alpha = 0.2) +
  labs(x = "Year", y = "Daily PM2.5 (µg/m³)", title = "Michigan daily PM2.5: 2005 vs 2025") +
  coord_cartesian(ylim = c(0, quantile(pm_mi$pm25, 0.99, na.rm = TRUE)))
```

Michigan daily PM2.5: 2005 vs 2025

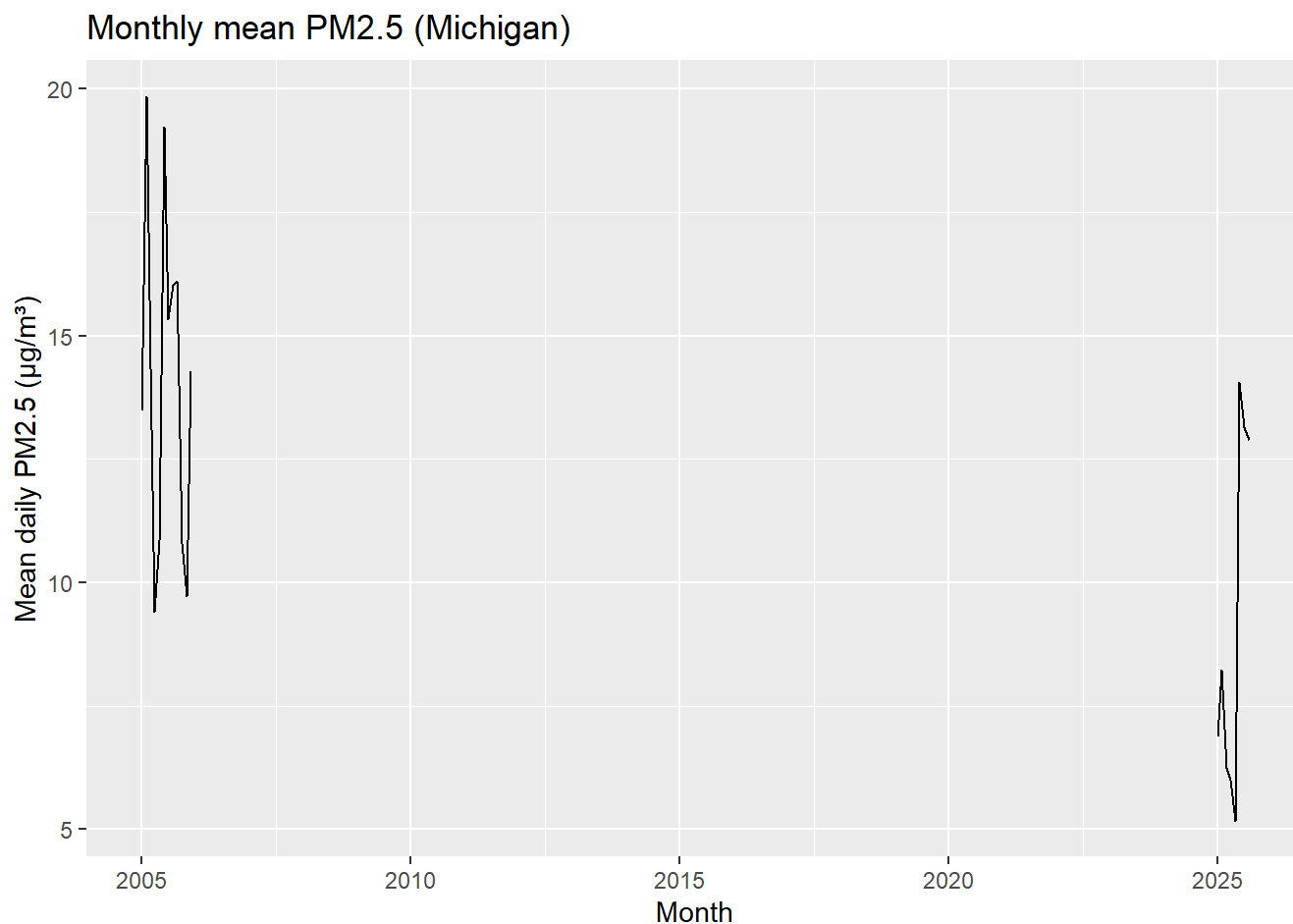


```
ggplot(pm_mi, aes(x = pm25)) +
  geom_histogram(bins = 60) +
  facet_wrap(~ year, ncol = 1) +
  labs(x = "Daily PM2.5 (µg/m³)", y = "Count", title = "Distribution of daily PM2.5 in Michigan")
  coord_cartesian(xlim = c(0, quantile(pm_mi$pm25, 0.99, na.rm = TRUE)))
```

Distribution of daily PM2.5 in Michigan



```
pm_mi_month <- pm_mi %>%  
  mutate(month = as.Date(format(Date, "%Y-%m-01"))) %>%  
  group_by(year, month) %>%  
  summarize(pm25_mean = mean(pm25, na.rm = TRUE), .groups = "drop")  
  
ggplot(pm_mi_month, aes(x = month, y = pm25_mean, group = factor(year))) +  
  geom_line() +  
  labs(x = "Month", y = "Mean daily PM2.5 (µg/m³)", title = "Monthly mean PM2.5 (Michigan)")
```



The summary statistics show that the mean and all of the percentiles (except for the max) of the daily PM2.5 concentration per month have decreased from 2005 to 2025; for example, the mean concentration decreased from 13.99 to 8.03 $\mu\text{g}/\text{m}^3$. The side-by-side box/violin plots and histograms also illustrate this well by showing a high frequency of values around 5 $\mu\text{g}/\text{m}^3$ in 2025, while concentrations of 20 $\mu\text{g}/\text{m}^3$ or higher (which were considered evenly distributed in 2005) are considered outliers in 2025. In general, the violin plot and line plot show that the distribution of concentrations has shifted down.

Level 2: PM2 concentrations by Michigan county

```
county_summary <- pm_mi %>%
  filter(!is.na(county)) %>%
  group_by(year, county) %>%
  summarize(
    n = sum(!is.na(pm25)),
    mean = mean(pm25, na.rm = TRUE),
    median = median(pm25, na.rm = TRUE),
    p95 = quantile(pm25, 0.95, na.rm = TRUE),
    .groups = "drop"
  )

county_summary
```

```
# A tibble: 37 × 6
  year county      n mean median  p95
<int> <chr>   <int> <dbl> <dbl> <dbl>
1  2005 Allegan    361 12.4   10    31
2  2005 Bay       117 12.4    9.7  28.2
3  2005 Berrien   118 13.1   10.8  31.3
4  2005 Chippewa  294  8.01   5.85  20.8
5  2005 Dickinson  25  7.32    6.9  13.2
6  2005 Genesee   119 12.9   10.7  27.4
7  2005 Ingham   120 13.5   11.2  30.0
8  2005 Iron      25  4.40    3.5   9.36
9  2005 Kalamazoo 167 14.1   12.6  30.6
10 2005 Kent      416 13.9    11   32.2
# i 27 more rows
```

```
county_change <- county_summary %>%
  select(year, county, mean, median, p95) %>%
  tidyr::pivot_wider(names_from = year, values_from = c(mean, median, p95)) %>%
  mutate(
    d_mean   = mean_2025 - mean_2005,
    d_median = median_2025 - median_2005,
    d_p95    = p95_2025 - p95_2005
  )

county_change %>% arrange(d_mean)
```

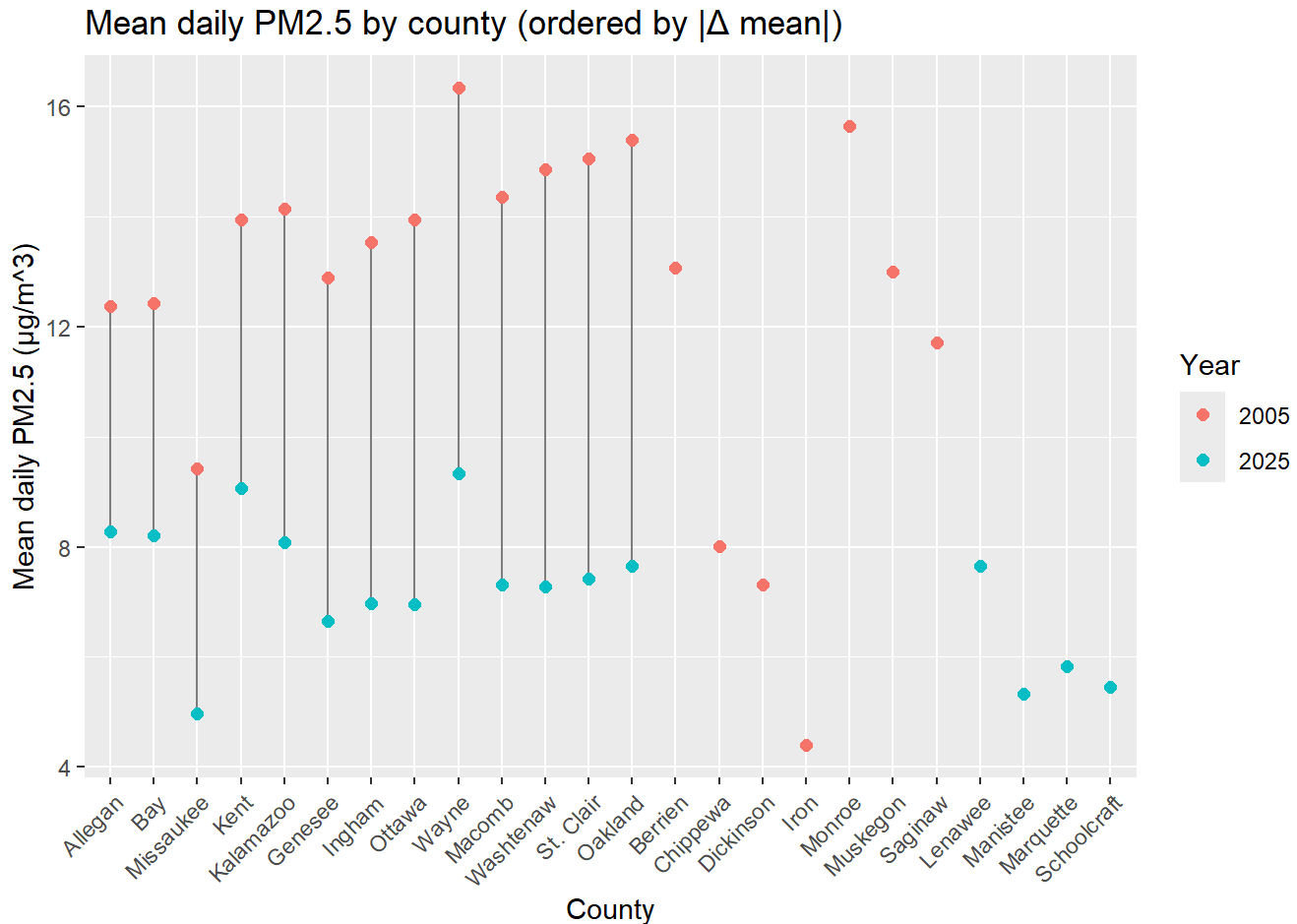
```
# A tibble: 24 × 10
  county mean_2005 mean_2025 median_2005 median_2025 p95_2005 p95_2025 d_mean
<chr>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
1 Oakland    15.4     7.66    12.3     5.76    36.1    19.7   -7.74
2 St. Cla...  15.0     7.43    12      5.72    36.6    17.8   -7.62
3 Washten...  14.9     7.28    11.8     5.57    33.3    17.5   -7.58
4 Macomb     14.4     7.32    10.8     5.43    32.6    17.5   -7.04
5 Wayne      16.3     9.33    13.5     7.2     35.6    22.6   -7.01
6 Ottawa     13.9     6.95    11.9     5.15    32.1    18.5   -6.99
7 Ingham     13.5     6.96    11.2     5.4     30.0    16.0   -6.57
8 Genesee    12.9     6.65    10.7     5.4     27.4    16.3   -6.23
9 Kalamaz...  14.1     8.07    12.6     6.22    30.6    19.6   -6.07
10 Kent       13.9     9.06    11      7.8     32.2    20.1   -4.88
# i 14 more rows
# i 2 more variables: d_median <dbl>, d_p95 <dbl>
```

```
county_levels <- county_change %>%
  arrange(abs(d_mean)) %>%
  pull(county)

plot_df <- county_summary %>%
  mutate(county = factor(county, levels = county_levels))

ggplot(plot_df, aes(x = county, y = mean, color = factor(year), group = county)) +
```

```
geom_line(color = "grey50") +
geom_point(size = 2) +
labs(x = "County", y = "Mean daily PM2.5 ( $\mu\text{g}/\text{m}^3$ )", color = "Year",
     title = "Mean daily PM2.5 by county (ordered by  $|\Delta \text{mean}|$ )") +
theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



From this plot, we can see that out of every county in Michigan where measurements were recorded in both 2005 and 2025, all of the measurements from 2025 are significantly smaller than those in 2005. According to the summary statistics, the median and 95th percentile also uniformly decrease from 2005 to 2025. The differences in concentrations range from -4.09 (in Allegan) to -7.74 $\mu\text{g}/\text{m}^3$ (in Oakland), meaning the concentrations decreased by at least 4 $\mu\text{g}/\text{m}^3$ and up to almost 8 $\mu\text{g}/\text{m}^3$ by county from 2005 to 2025.

We also see an overall decrease when comparing the 7 measurements for counties only taken in 2005 compared to the 4 measurements for counties only taken in 2025, since none of the measurements from counties only recorded in 2025 are higher than 8 $\mu\text{g}/\text{m}^3$.

```
library(maps)

mi_map <- map_data("county") %>%
  filter(region == "michigan") %>%
  mutate(county = str_to_title(subregion))

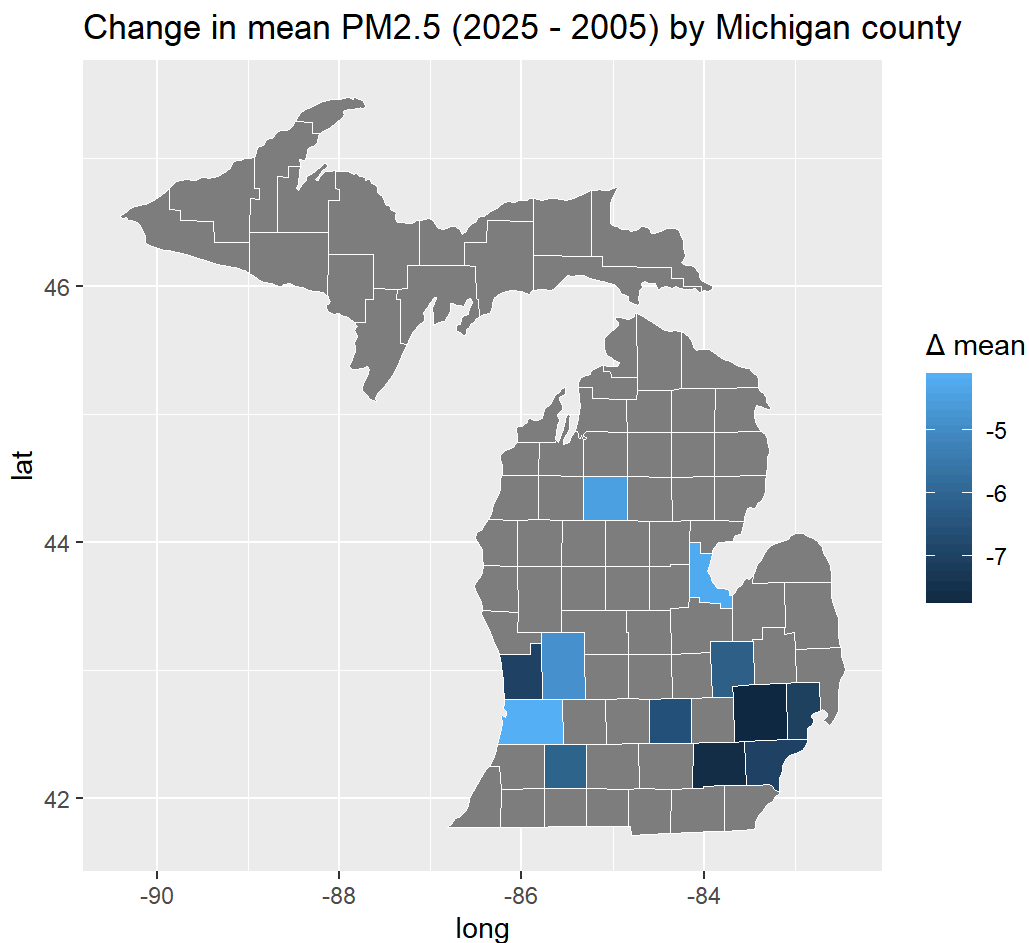
county_map_df <- mi_map %>%
```

```

left_join(county_change, by = "county")

ggplot(county_map_df, aes(long, lat, group = group, fill = d_mean)) +
  geom_polygon(color = "white", linewidth = 0.2) +
  coord_quickmap() +
  labs(title = "Change in mean PM2.5 (2025 - 2005) by Michigan county",
       fill = "Δ mean")

```



This choropleth shows the changes in mean PM2.5 concentrations per Michigan county from 2005 to 2025. We can see that the largest decreases (the deepest blue colors) are in counties in the southeast region of Michigan, which (as discussed previously) is where Metro Detroit is located. This suggests that the decreases are concentrated in metro/industrial areas, and the decreases are less pronounced in more rural counties closer to the north side of Michigan.

Level 3: PM2 concentrations by site in Wayne County

```

pm_wayne <- pm_mi %>%
  filter(county == "Wayne") %>%
  mutate(site_id = paste0(county, "-", site))

site_summary <- pm_wayne %>%
  group_by(year, site_id, city_name) %>%

```



```

summarize(
  n = sum(!is.na(pm25)),
  mean = mean(pm25, na.rm = TRUE),
  median = median(pm25, na.rm = TRUE),
  p95 = quantile(pm25, 0.95, na.rm = TRUE),
  .groups = "drop"
)

```

```
site_summary
```

```
# A tibble: 18 × 7
```

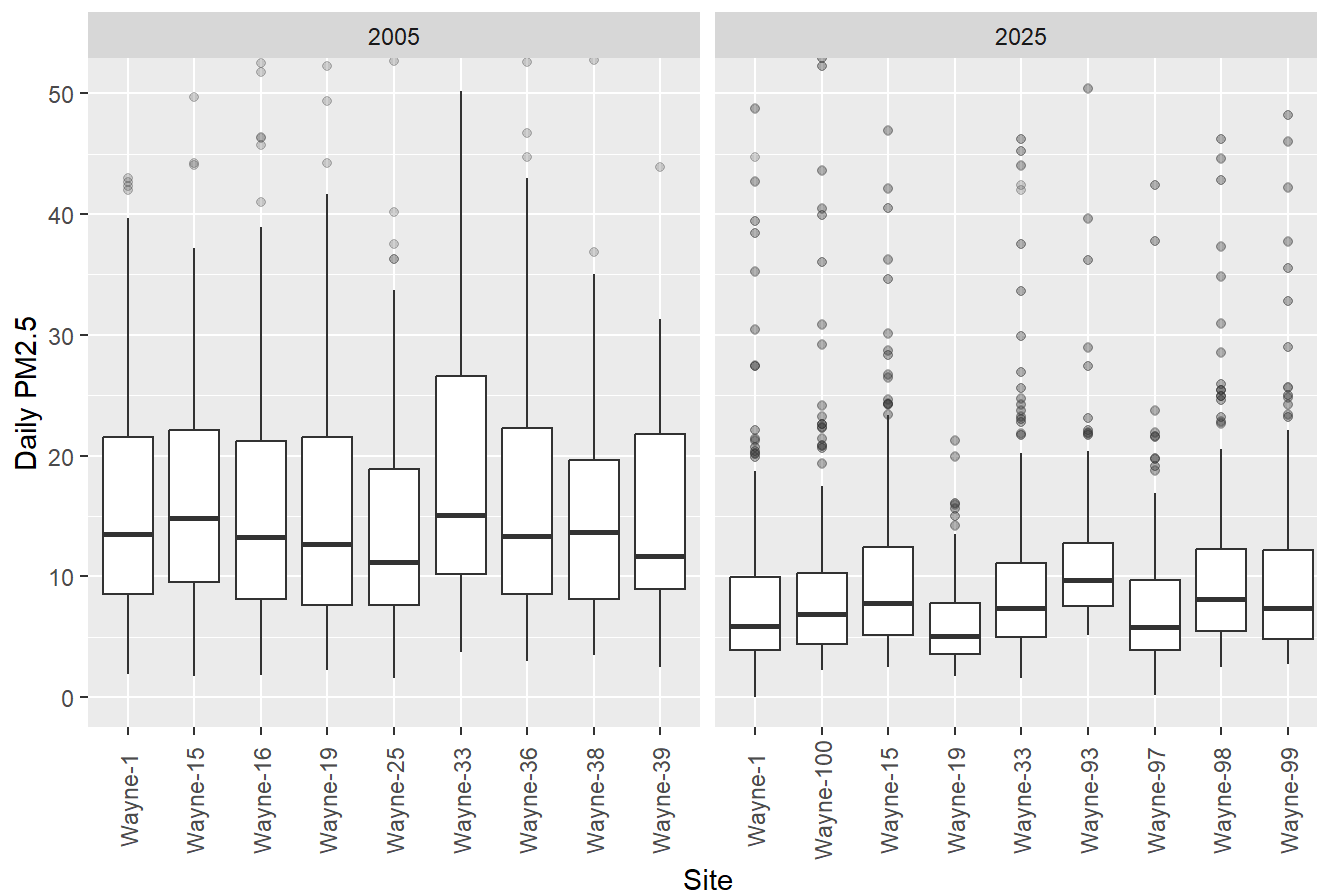
	year	site_id	city_name	n	mean	median	p95
	<int>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>
1	2005	Wayne-1	<NA>	408	16.2	13.5	33.7
2	2005	Wayne-15	<NA>	114	17.2	14.8	34.2
3	2005	Wayne-16	<NA>	338	16.0	13.2	35.4
4	2005	Wayne-19	<NA>	117	16.4	12.7	41.5
5	2005	Wayne-25	<NA>	114	14.9	11.2	34.6
6	2005	Wayne-33	<NA>	115	18.5	15.1	36.2
7	2005	Wayne-36	<NA>	113	16.4	13.3	34.2
8	2005	Wayne-38	<NA>	75	16.4	13.7	35.6
9	2005	Wayne-39	<NA>	35	15.0	11.7	28.8
10	2025	Wayne-1	<NA>	533	8.32	5.91	20.5
11	2025	Wayne-100	<NA>	486	8.97	6.86	22.4
12	2025	Wayne-15	<NA>	486	10.3	7.8	24.7
13	2025	Wayne-19	<NA>	282	6.27	5.08	13.5
14	2025	Wayne-33	<NA>	555	9.71	7.34	23.4
15	2025	Wayne-93	<NA>	330	11.3	9.7	21.8
16	2025	Wayne-97	<NA>	362	7.67	5.81	19.2
17	2025	Wayne-98	<NA>	486	10.4	8.1	24.9
18	2025	Wayne-99	<NA>	428	10.1	7.34	25.0

```

ggplot(pm_wayne, aes(x = site_id, y = pm25)) +
  geom_boxplot(outlier.alpha = 0.2) +
  facet_wrap(~ year, scales = "free_x") +
  labs(x = "Site", y = "Daily PM2.5", title = "Wayne County: daily PM2.5 by site") +
  coord_cartesian(ylim = c(0, quantile(pm_wayne$pm25, 0.99, na.rm = TRUE))) +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5))

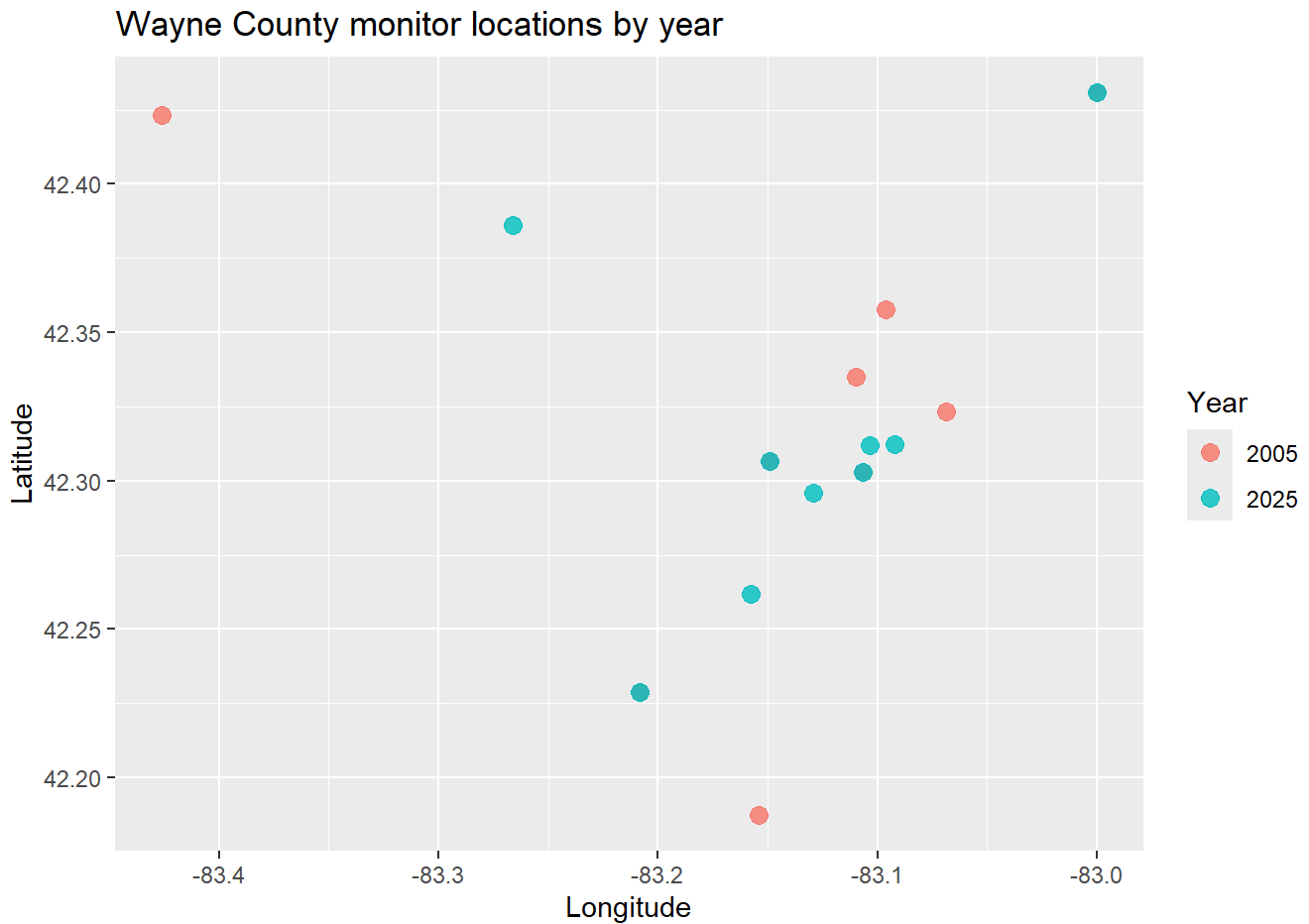
```

Wayne County: daily PM2.5 by site



```
wayne_sites <- pm_wayne %>%
  filter(!is.na(lat), !is.na(lon)) %>%
  distinct(year, site_id, city_name, lat, lon)

ggplot(wayne_sites, aes(x = lon, y = lat, color = factor(year))) +
  geom_point(size = 3, alpha = 0.8) +
  labs(x = "Longitude", y = "Latitude", color = "Year",
       title = "Wayne County monitor locations by year") +
  coord_quickmap()
```



From the series of boxplots that show the distribution of daily PM_{2.5} concentrations in all of the Wayne County sites, we see that the medians have uniformly decreased from 2005 to 2025; in fact, according to the summary statistics, the lowest mean and median concentrations in 2005 are still higher than the highest mean and median concentrations in 2025. The boxplots also show (much like the broad, state-wide analysis from before) that the concentrations above 20 $\mu\text{g}/\text{m}^3$ are within the 75th percentile in 2005, but outliers in 2025. Finally, the spatial plot of site locations colored by year shows that the sites in 2005 (red points) are more spread out across Wayne County, but in 2025 (blue points), they are more clustered in one area between latitude 42.30 and longitude -83.15, with only three sites located away from that cluster.

Therefore, we can finally conclude that on the level of the entire state of Michigan, on the level of each county within Michigan, and the level of each observation site within Wayne County, the daily concentrations of PM_{2.5} have decreased in Michigan over the 20 years spanning from 2005 to 2025.