

# Tidy Tuesday

Week 33

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PUBLISHED  
August 20, 2024

```
# Load the tidyuesday package
suppressMessages(library(tidyuesdayR))
suppressMessages(library(skimr))
suppressMessages(library(tidyverse))
suppressMessages(library(dplyr))

# Load the current week's dataset
tuesdata <- tidyuesdayR::tt_load('2024-08-13')
```

Downloading file 1 of 1: `worlds\_fairs.csv`

```
worlds_fairs <- tuesdata$worlds_fairs

# Explore the structure of the dataset
str(worlds_fairs)
```

```
spc_tbl_ [70 × 14] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
 $ start_month      : num [1:70] 4 5 5 4 5 5 5 10 4 5 ...
 $ start_year       : num [1:70] 1851 1855 1862 1867 1873 ...
 $ end_month        : num [1:70] 10 11 11 11 10 11 11 4 12 10 ...
 $ end_year         : num [1:70] 1851 1855 1862 1867 1873 ...
 $ name_of_exposition : chr [1:70] "The Great Exhibition" "Exposition Universelle / Paris
International" "International Exhibition" "Exposition Universelle / Paris International"
...
 $ country          : chr [1:70] "United Kingdom" "France" "United Kingdom" "France"
...
 $ city             : chr [1:70] "London" "Paris" "London" "Paris" ...
 $ category         : chr [1:70] "World Expo" "World Expo" "World Expo" "World Expo"
...
 $ theme            : chr [1:70] "Industry of all Nations" "Agriculture, Industry and
Arts" "Industry and Arts" "Agriculture, Industry and Arts" ...
 $ notables         : chr [1:70] "The Crystal Palace" "Palais d'Industrie, Bordeaux
Wine classification" NA "Champ de Mars" ...
 $ visitors         : num [1:70] 6 5 6 15 7.25 10 16 1.3 2.3 32 ...
 $ cost             : num [1:70] 165 2 2 45 95 8 11 16 17 3 ...
 $ area             : num [1:70] 10 15 15 69 233 115 75 25 47 96 ...
 $ attending_countries: num [1:70] 25 25 39 42 35 35 36 33 30 35 ...
- attr(*, "spec")=
 .. cols(
```

```

.. start_month = col_double(),
.. start_year = col_double(),
.. end_month = col_double(),
.. end_year = col_double(),
.. name_of_exposition = col_character(),
.. country = col_character(),
.. city = col_character(),
.. category = col_character(),
.. theme = col_character(),
.. notables = col_character(),
.. visitors = col_double(),
.. cost = col_double(),
.. area = col_double(),
.. attending_countries = col_double()
.. )
- attr(*, "problems")=<externalptr>

```

```
skim(worlds_fairs)
```

Name	worlds_fairs
Number of rows	70
Number of columns	14

---

Column type frequency:

character	6
numeric	8

---

Group variables	None
-----------------	------

Data summary

## Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
name_of_exposition	0	1.00	5	101	0	67	0
country	0	1.00	5	26	0	24	0
city	0	1.00	4	14	0	51	0
category	0	1.00	10	16	0	2	0
theme	0	1.00	5	77	0	68	0
notables	11	0.84	3	83	0	58	0

## Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
start_month	0	1.00	5.41	1.88	2.00	4.00	5.0	6.00	12.00	
start_year	0	1.00	1947.19	44.26	1851.00	1911.50	1953.5	1983.50	2021.00	
end_month	0	1.00	9.16	2.38	1.00	9.00	10.0	11.00	12.00	
end_year	0	1.00	1947.29	44.26	1851.00	1911.50	1953.5	1983.50	2022.00	
visitors	14	0.80	16.64	16.59	0.80	5.64	10.0	21.63	73.08	
cost	34	0.51	559.94	998.43	2.00	26.75	125.5	448.25	4200.00	
area	5	0.93	105.82	116.99	0.08	30.00	70.0	120.00	523.00	
attending_countries	5	0.93	49.22	45.04	8.00	23.00	33.0	55.00	192.00	

```
# View the first few rows of the dataset
head(worlds_fairs)
```

start_month	start_year	end_month	end_year
<dbl>	<dbl>	<dbl>	<dbl>
4	1851	10	1851
5	1855	11	1855
5	1862	11	1862
4	1867	11	1867
5	1873	10	1873
5	1876	11	1876

6 rows | 1-4 of 14 columns

```
#tidytuesdayR::use_tidytemplate()
```

## Missing Values

Notables: 11 missing values (about 15.7% of the rows). Visitors: 14 missing values (about 20% of the rows). Cost: 34 missing values (about 48.6% of the rows). Area: 5 missing values (about 7.1% of the rows). Attending Countries: 5 missing values (about 7.1% of the rows). Most of the cost fetched from <https://jdpecon.com/expo/expolist.html>

## Clean the data

```
#selected_data <- worlds_fairs %>% dplyr::select("country", "start_year", "city",
#selected_data

# add missing values
# https://jdpecon.com/expo/expolist.html
worlds_fairs[70, "cost"] <- 14600 # Dubai 2022
worlds_fairs[69, "cost"] <- 3000 # Kazakhstan 2017
```

```

worlds_fairs[68, "cost"] <- 1640 # Italy 2015
worlds_fairs[65, "cost"] <- 1070 # Spain 2008
worlds_fairs[64, "cost"] <- 3130 # Japan 2005
worlds_fairs[62, "cost"] <- 2952 # Portugal 1998 | Estimated $1753 to $2398 milli
worlds_fairs[61, "cost"] <- 2800 # South Korea 1993
worlds_fairs[60, "cost"] <- 568.75 # Italy 1992 $487.5 to $650 million according
worlds_fairs[59, "cost"] <- 2332 # Spain 1992
worlds_fairs[54, "cost"] <- 763.7 # Japan 1985
worlds_fairs[50, "cost"] <- 304 # Japan 1975
worlds_fairs[12, "cost"] <- 1.14 # Belgium 1897

# Corrections
# https://jdpecon.com/expo/expolist.html
worlds_fairs[1, "cost"] <- 1.67 # London 1851
worlds_fairs[2, "cost"] <- 5.7 # France 1855
worlds_fairs[3, "cost"] <- 2.3 # London 1862
worlds_fairs[4, "cost"] <- 4.5 # Paris 1867
worlds_fairs[5, "cost"] <- 12 # Vienna 1873
worlds_fairs[6, "cost"] <- 9 # Philadelphia 1876
# worlds_fairs[7, "cost"] <- 11 # France 1878
worlds_fairs[8, "cost"] <- 1.56 # Melbourne 1880
worlds_fairs[9, "cost"] <- 1.7 # Spain 1881
worlds_fairs[10, "cost"] <- 8.01 # France 1889 41.5 million French francs would b
worlds_fairs[11, "cost"] <- 27.3 # Chicago 1893
worlds_fairs[12, "cost"] <- 1.14 # Belgium 1897
worlds_fairs[13, "cost"] <- 18.75 # Paris 1900
worlds_fairs[14, "cost"] <- 26.5 # St Louis 1904
worlds_fairs[15, "cost"] <- 2.89 # Belgium 1905
worlds_fairs[16, "cost"] <- 2.6 # Milan 1906
worlds_fairs[17, "cost"] <- 3.55 # Belgium 1910
# worlds_fairs[18, "cost"] <- NA # Italy 1911
worlds_fairs[19, "cost"] <- 3.3 # Belgium 1913
worlds_fairs[20, "cost"] <- 24.7 # San Francisco 1915
# worlds_fairs[21, "cost"] <- 25 # Spain 1929
worlds_fairs[22, "cost"] <- 31 # Chicago 1933
worlds_fairs[23, "cost"] <- 22.4 # Belgium 1935
# worlds_fairs[24, "cost"] <- NA # Sweden 1936
worlds_fairs[25, "cost"] <- 57.25 # Paris 1937
# worlds_fairs[26, "visitors"] <- NA # Finland 1938
worlds_fairs[27, "cost"] <- 160 # New York City 1939
worlds_fairs[27, "attending_countries"] <- 52 # New York City 1939

# https://en.worldfairs.info/expodonnees.php?expo_id=32
worlds_fairs[28, "area"] <- 80 # Belgium 1939
worlds_fairs[28, "visitors"] <- 1.7 # Belgium 1939
worlds_fairs[28, "cost"] <- 58.51 # Belgium 1939 | 150 million Belgian francs wo

# worlds_fairs[29, "cost"] <- NA # France 1947
# worlds_fairs[30, "cost"] <- NA # Sweden 1949
# worlds_fairs[31, "cost"] <- NA # France 1949

```

```

# https://www.bie-paris.org/site/en/latest/blog/entry/cultivating-tourism-in-haiti
worlds_fairs[32, "cost"] <- 4 # Expo Port-au-Prince 1949
worlds_fairs[32, "visitors"] <- 0.07 # Expo Port-au-Prince 1949
# worlds_fairs[33, "cost"] <- NA # France 1951
# worlds_fairs[34, "cost"] <- NA # Italy 1953
worlds_fairs[35, "visitors"] <- 0.6 # Expo 1953 Jerusalem
# worlds_fairs[36, "cost"] <- NA # Italy 1954
worlds_fairs[37, "visitors"] <- 0.12 # Italy 1955
# worlds_fairs[38, "cost"] <- NA # Sweden 1955
# worlds_fairs[39, "cost"] <- NA # Israel 1956
# worlds_fairs[40, "cost"] <- NA # West Germany 1957
worlds_fairs[41, "cost"] <- 43.4 # Belgium 1958
# worlds_fairs[42, "cost"] <- NA # Italy 1961
worlds_fairs[43, "cost"] <- 22.8 # Seattle 1962
# worlds_fairs[44, "cost"] <- NA # West Germany 1965
worlds_fairs[45, "cost"] <- 384.7 # 1967 Canada 415.920 CAD million would be approx
# worlds_fairs[46, "cost"] <- 156.00 # San Antonio 1968
worlds_fairs[47, "cost"] <- 247.54 # Japan 1970
# worlds_fairs[48, "cost"] <- NA # Hungary 1971
worlds_fairs[49, "cost"] <- 78.40 # Spokane 1974
worlds_fairs[50, "cost"] <- 155.2 # Japan 1975
# worlds_fairs[51, "cost"] <- NA # Bulgaria 1981
worlds_fairs[52, "cost"] <- 111 # Knoxville 1982
worlds_fairs[53, "cost"] <- 442.5 # New Orleans 1984
worlds_fairs[54, "cost"] <- 763.7 # Japan 1985
# worlds_fairs[55, "cost"] <- NA # Bulgaria 1985
worlds_fairs[56, "cost"] <- 609.52 # Canada 1986
worlds_fairs[57, "cost"] <- 642.2 # Australia 1988
# worlds_fairs[58, "cost"] <- NA # Bulgaria 1991
# worlds_fairs[59, "cost"] <- 2332 # Spain 1992
# worlds_fairs[60, "cost"] <- 568.75 # Italy 1992
# worlds_fairs[61, "cost"] <- 2800 # South Korea 1993
# worlds_fairs[62, "cost"] <- 2952 # Portugal 1998
worlds_fairs[63, "cost"] <- 2238 # Germany 2000
# worlds_fairs[64, "cost"] <- 3130 # Japan 2005
# worlds_fairs[65, "cost"] <- 1070 # Spain 2008
worlds_fairs[66, "cost"] <- 8800 # China 2010

#https://www.exhibitionworld.co.uk/crossing-oceans-expo-2012-yeosu-korea
worlds_fairs[67, "cost"] <- 2000 # South Korea 2012

worlds_fairs[68, "cost"] <- 1640 # Italy 2015
worlds_fairs[69, "cost"] <- 3000 # Kazakhstan 2017
worlds_fairs[70, "cost"] <- 14600 # Dubai 2022

# we still have 21 null values for cost

```

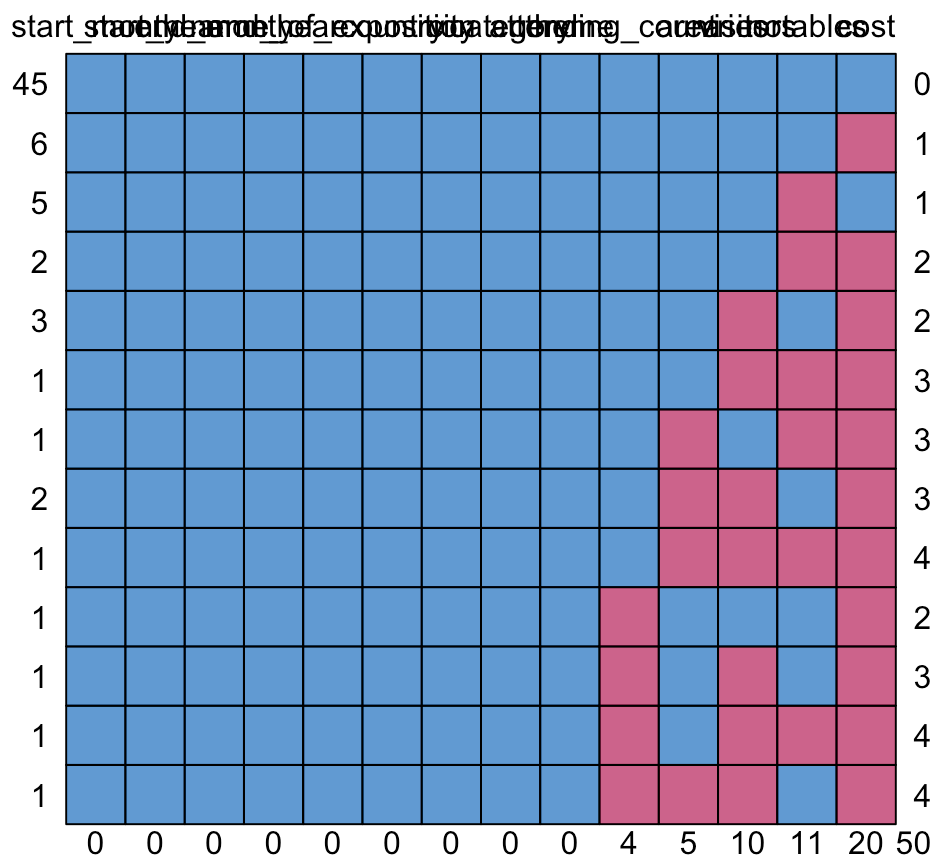
```
# visitors 10
# area 4 and attending_countries 5
```

To address the missing values, we will use an imputation method to help fill in the remaining missing values using a more sophisticated technique than the median. Using the mice package (<https://amices.org/mice/>), we will consider the relationships between variables to fill in the missing data.

Instead of using KNN, we decided to try PMM (Predictive Mean Matching). This matches each missing value with the nearest observed values and imputes from those values. Effective for maintaining the distribution of the data. PMM does multiple imputations by default when you set  $m > 1$ , creating multiple versions of the dataset with imputed values, reflecting the uncertainty of the missing data.

```
suppressMessages(library(mice))

# Show the missing data pattern
md.pattern(worlds_fairs) # Blue is observed, red is missing
```



	start_month	start_year	end_month	end_year	name_of_exposition	country	city
45	1	1	1	1		1	1
6	1	1	1	1		1	1
5	1	1	1	1		1	1
2	1	1	1	1		1	1
3	1	1	1	1		1	1

1	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1
2	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1
0	0	0	0	0	0	0	0
category theme attending_countries area visitors notables cost							
45	1	1	1	1	1	1	0
6	1	1	1	1	1	1	1
5	1	1	1	1	1	0	1
2	1	1	1	1	1	0	2
3	1	1	1	1	0	1	2
1	1	1	1	1	0	0	3
1	1	1	1	0	1	0	3
2	1	1	1	0	0	1	3
1	1	1	1	0	0	0	4
1	1	1	0	1	1	1	2
1	1	1	0	1	0	1	3
1	1	1	0	1	0	0	4
1	1	1	0	0	0	1	4
0	0		4	5	10	11	50

```
imputed_data <- mice(worlds_fairs, method = 'pmm', m = 5, maxit = 50)
```

iter	imp	variable
1	1	visitors cost area attending_countries
1	2	visitors cost area attending_countries
1	3	visitors cost area attending_countries
1	4	visitors cost area attending_countries
1	5	visitors cost area attending_countries
2	1	visitors cost area attending_countries
2	2	visitors cost area attending_countries
2	3	visitors cost area attending_countries
2	4	visitors cost area attending_countries
2	5	visitors cost area attending_countries
3	1	visitors cost area attending_countries
3	2	visitors cost area attending_countries
3	3	visitors cost area attending_countries
3	4	visitors cost area attending_countries
3	5	visitors cost area attending_countries
4	1	visitors cost area attending_countries
4	2	visitors cost area attending_countries
4	3	visitors cost area attending_countries
4	4	visitors cost area attending_countries
4	5	visitors cost area attending_countries
5	1	visitors cost area attending_countries

[illegible]



[illegible]

[illegible]

[illegible]

```

46 1 visitors cost area attending_countries
46 2 visitors cost area attending_countries
46 3 visitors cost area attending_countries
46 4 visitors cost area attending_countries
46 5 visitors cost area attending_countries
47 1 visitors cost area attending_countries
47 2 visitors cost area attending_countries
47 3 visitors cost area attending_countries
47 4 visitors cost area attending_countries
47 5 visitors cost area attending_countries
48 1 visitors cost area attending_countries
48 2 visitors cost area attending_countries
48 3 visitors cost area attending_countries
48 4 visitors cost area attending_countries
48 5 visitors cost area attending_countries
49 1 visitors cost area attending_countries
49 2 visitors cost area attending_countries
49 3 visitors cost area attending_countries
49 4 visitors cost area attending_countries
49 5 visitors cost area attending_countries
50 1 visitors cost area attending_countries
50 2 visitors cost area attending_countries
50 3 visitors cost area attending_countries
50 4 visitors cost area attending_countries
50 5 visitors cost area attending_countries

```

```

#imputed_data <- mice(worlds_fairs, method = 'norm', m = 5, maxit = 50) # Gives r
#imputed_data <- mice(worlds_fairs, method = 'norm.predict', m = 5, maxit = 50) #
#imputed_data <- mice(worlds_fairs, method = 'mean', m = 5, maxit = 50) # Gives r
#imputed_data <- mice(worlds_fairs, method = 'pmm', m = 5, maxit = 50, seed = 184
# we are using the Predictive Mean Matching method and we generated 5 different v

# Inspect the imputed data.
#summary(imputed_data)

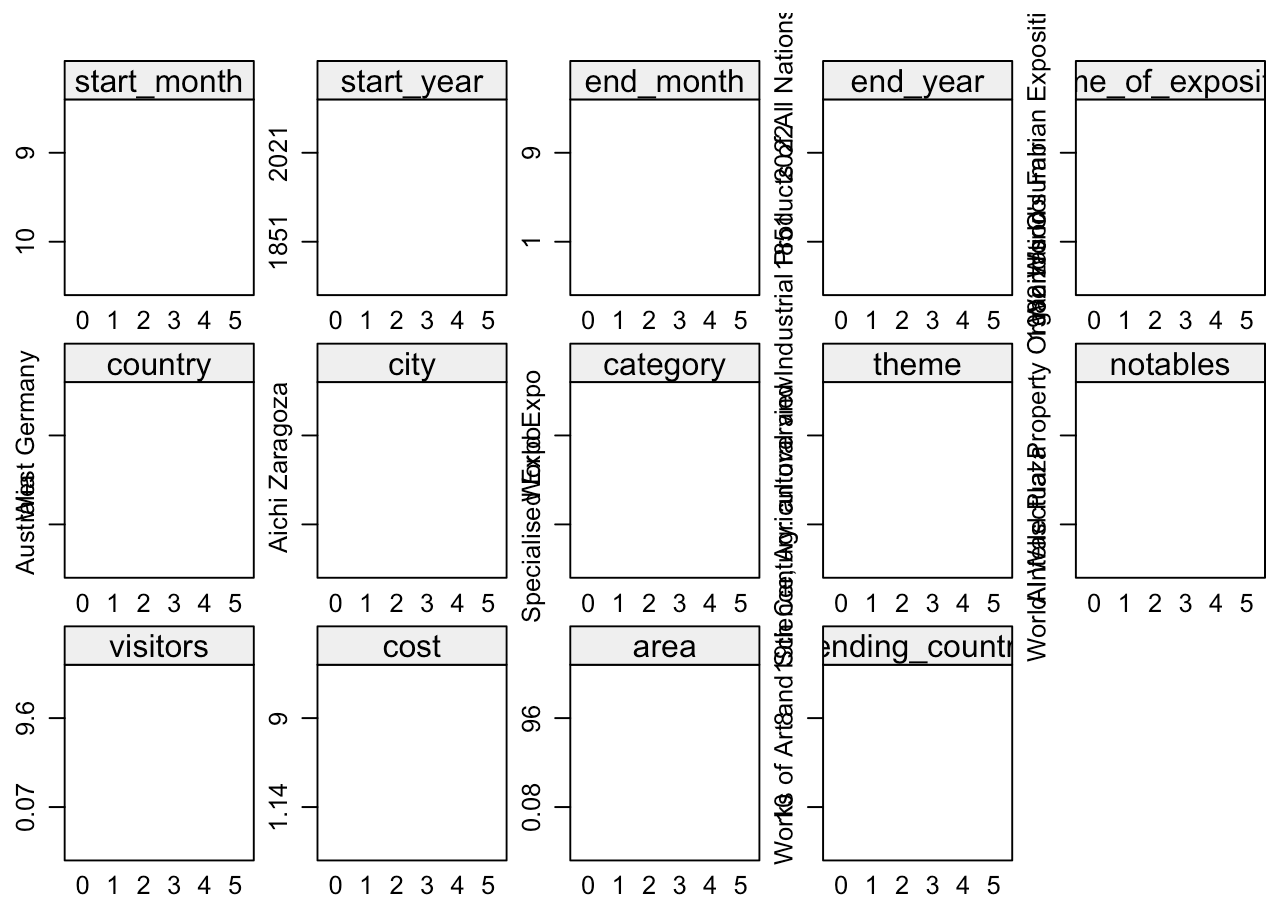
# Matrix of the imputed values for the cost variable
imputed_data$imp$cost

```

	1	2	3	4	5
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
18	442.50	1.14	3.55	160.00	22.80
24	1.70	111.00	3.55	3.55	78.40
26	1.14	442.50	24.70	22.40	1.70
29	3.55	57.25	24.70	2.30	4.50
30	609.52	4.00	12.00	43.40	568.75
31	12.00	3130.00	25.00	12.00	3130.00
33	160.00	5.70	8.01	160.00	5.70
34	58.51	25.00	27.30	642.20	2.89

	1	2	3	4	5
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
35	58.51	1070.00	763.70	8.01	27.30
36	22.40	609.52	8.01	57.25	156.00
1-10 of 20 rows				Previous	1 2 Next

```
# Visual summary of the imputation process
stripplot(imputed_data)
```



```
# helps you understand how the imputed values compare to the observed data and wh

# Extract the Dataset
worlds_fairs_imputed <- complete(imputed_data,3)

# Compare Before and After
summary(worlds_fairs$cost)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
1.140	8.258	50.325	986.281	634.030	14600.000	20

```
summary(worlds_fairs_imputed$cost)
```

```
Min.   1st Qu.   Median     Mean 3rd Qu.    Max.
1.14    8.01    25.00   758.60  537.19 14600.00
```

```
#skim(worlds_fairs)
#skim(worlds_fairs_imputed)

imputed_fairs <- worlds_fairs_imputed %>% dplyr::select("start_year", "country", "
imputed_fairs
```

start_year	country	city	cost
<dbl>	<chr>	<chr>	<dbl>
1851	United Kingdom	London	1.67
1855	France	Paris	5.70
1862	United Kingdom	London	2.30
1867	France	Paris	4.50
1873	Austria-Hungary	Vienna	12.00
1876	United States	Philadelphia	9.00
1878	France	Paris	11.00
1880	Colony of Victoria	Melbourne	1.56
1888	Spain	Barcelona	1.70
1889	France	Paris	8.01

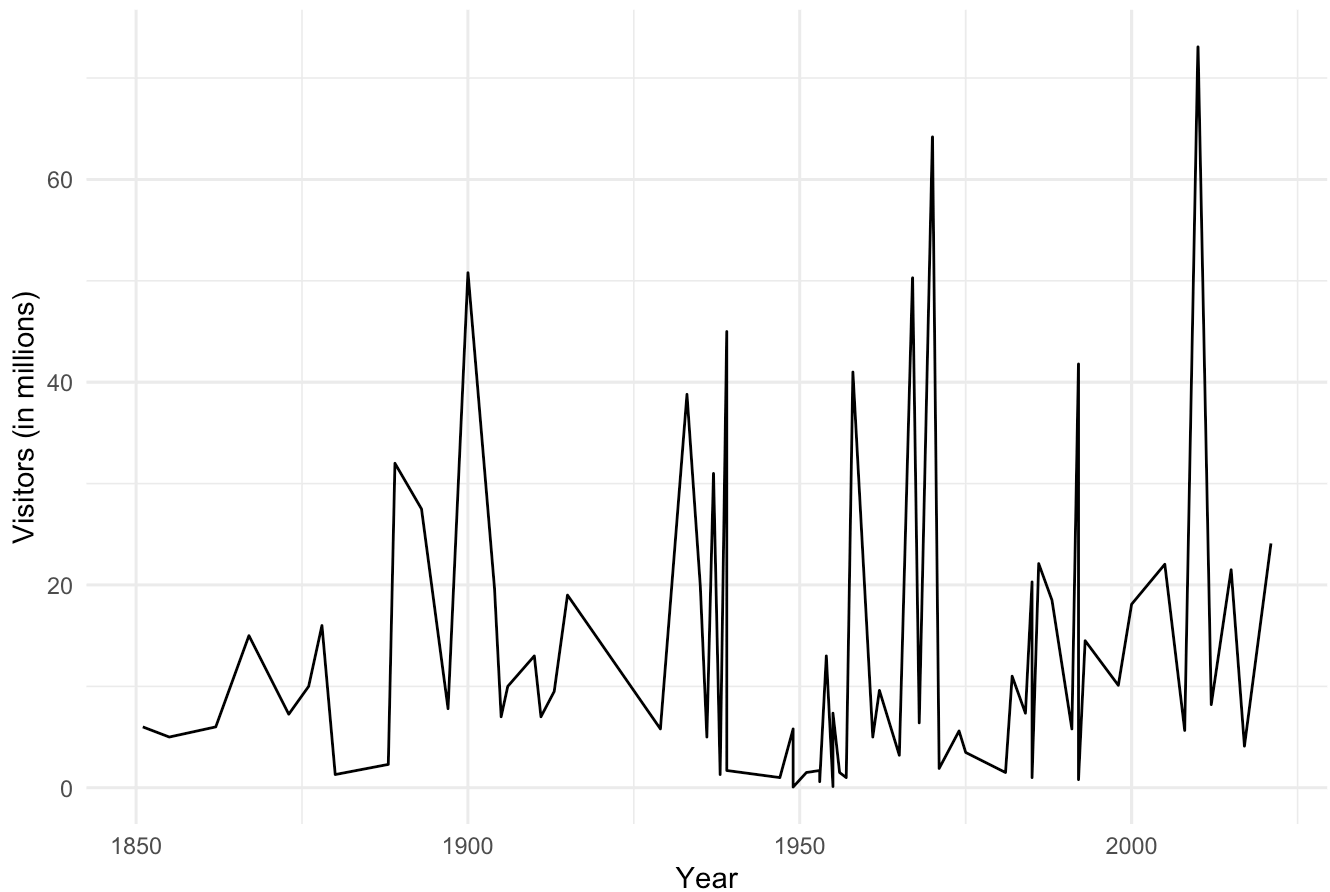
1-10 of 70 rows

Previous1234567Next

```
library(ggplot2)

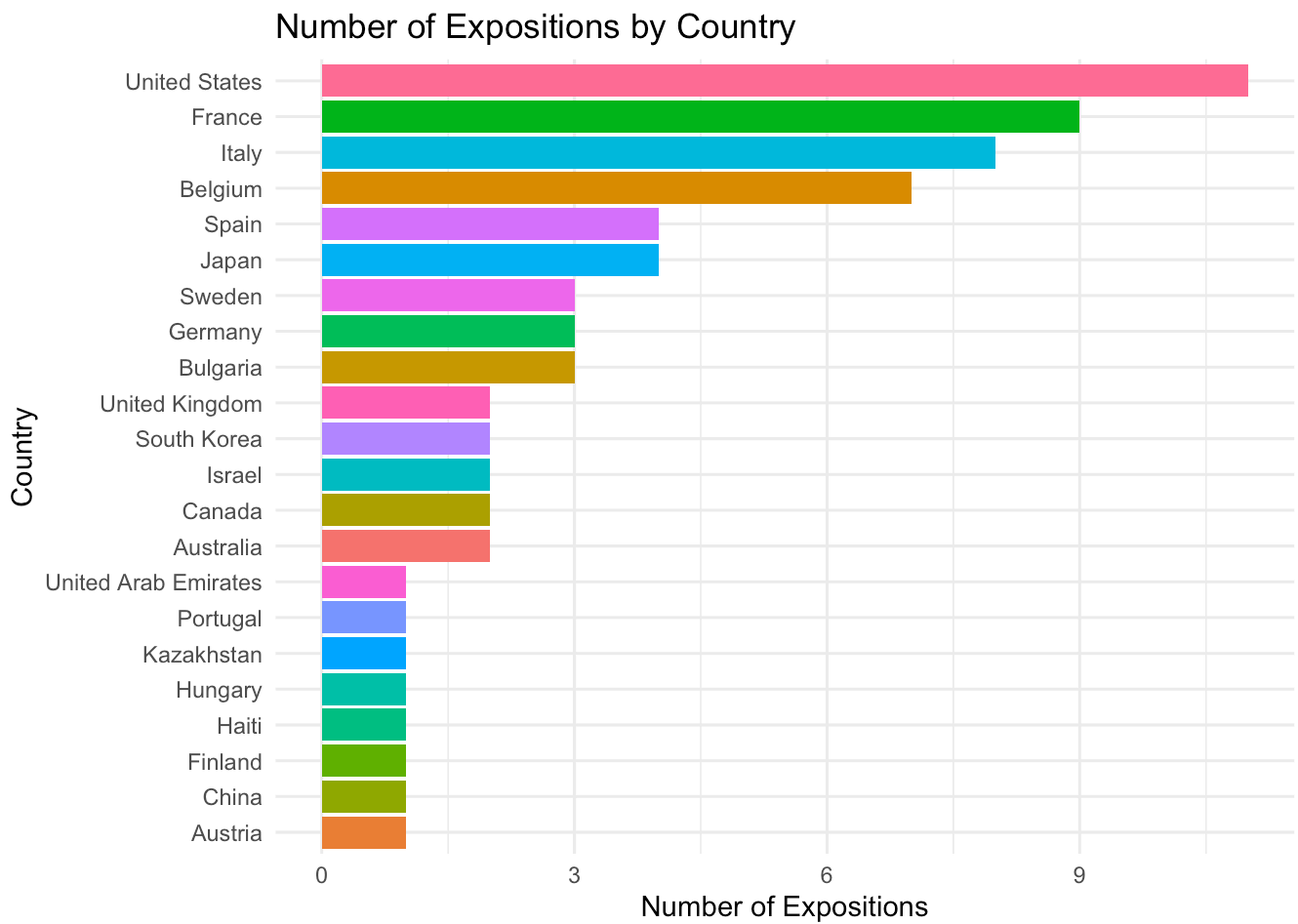
ggplot(worlds_fairs_imputed, aes(x = start_year, y = visitors)) +
  geom_line() +
  labs(title = "Visitor Trends Over Time",
       x = "Year",
       y = "Visitors (in millions)") +
  theme_minimal()
```

## Visitor Trends Over Time



```
# Rename countries to
worlds_fairs_imputed <- worlds_fairs_imputed %>%
  mutate(country = recode(country, "Colony of Victoria" = "Australia", "West Germ

ggplot(worlds_fairs_imputed %>%
  count(country) %>%
  arrange(n),
  aes(x = reorder(country, n), y = n, fill = country)) +
geom_bar(stat = "identity") +
coord_flip() +
labs(title = "Number of Expositions by Country",
  x = "Country",
  y = "Number of Expositions") +
theme_minimal() +
theme(legend.position = "none")
```



```
cost_expositions_df <- worlds_fairs_imputed[, c("cost", "start_year")]

cost_expositions_df
```

	cost	start_year
	<dbl>	<dbl>
	1.67	1851
	5.70	1855
	2.30	1862
	4.50	1867
	12.00	1873
	9.00	1876
	11.00	1878
	1.56	1880
	1.70	1888
	8.01	1889

1-10 of 70 rows

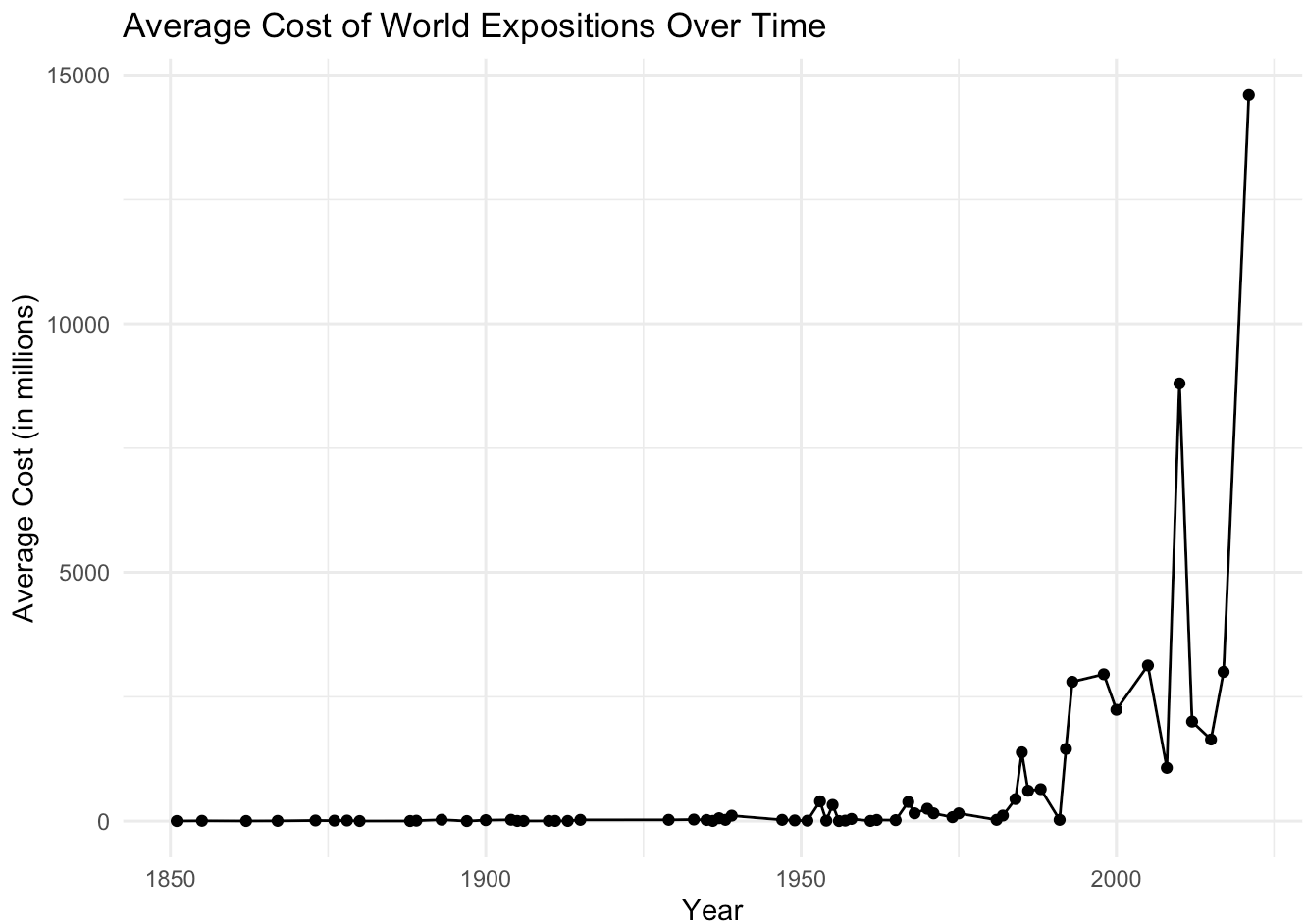
Previous1234567Next



```

worlds_fairs_imputed %>%
  group_by(start_year) %>%
  summarise(avg_cost = mean(cost, na.rm = TRUE)) %>%
  ggplot(aes(x = start_year, y = avg_cost)) +
  geom_line() +
  geom_point() +
  labs(title = "Average Cost of World Expositions Over Time",
       x = "Year",
       y = "Average Cost (in millions)") +
  theme_minimal()

```



```

#sessioninfo::session_info(include_base = TRUE)

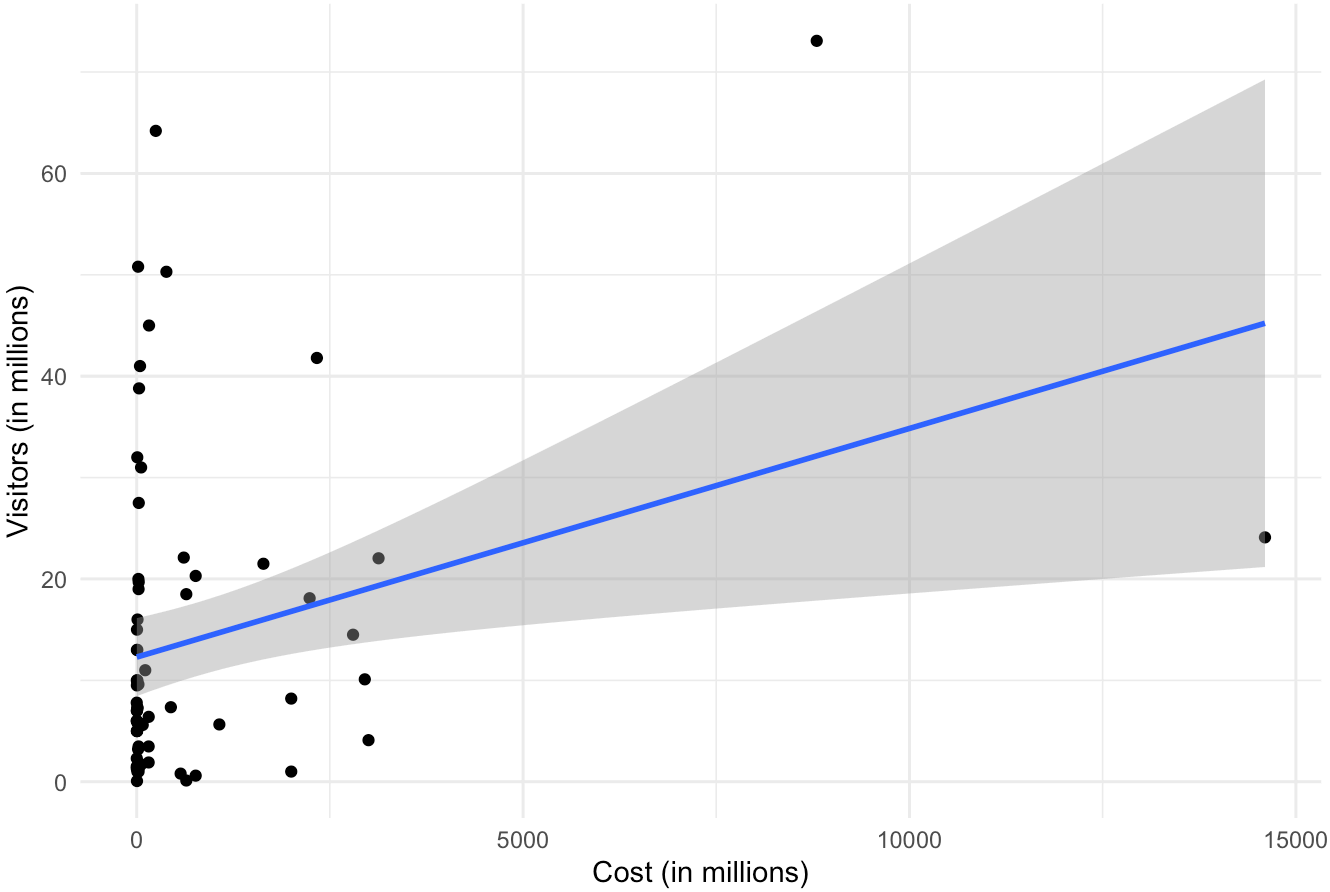
```

```

ggplot(worlds_fairs_imputed, aes(x = cost, y = visitors)) +
  geom_point() +
  geom_smooth(method = "lm") +
  labs(title = "Cost vs. Visitors",
       x = "Cost (in millions)",
       y = "Visitors (in millions)") +
  theme_minimal()

```

Cost vs. Visitors



```
cost_visitors_df <- worlds_fairs_imputed[, c("cost", "visitors")]

cost_visitors_df
```

cost	visitors
<dbl>	<dbl>
1.67	6.000
5.70	5.000
2.30	6.000
4.50	15.000
12.00	7.250
9.00	10.000
11.00	16.000
1.56	1.300
1.70	2.300
8.01	32.000

