

DallasArrests

Group Project

2024-11-03

Introduction

This report analyzes the relationship between arrest rates and property prices in Dallas by zip code. Using datasets on arrests and property values, we aim to identify any correlations that may exist at the neighborhood level.

Data handling

```
library(ggplot2)
# Load necessary libraries
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

# Load the data
arrests <- read.csv("DallasArrests_cleaned.csv")
zillow <- read.csv("Zillow-dallas.csv")

# Count Arrests per Zip Code
arrests_per_zip <- arrests %>%
  group_by(Zipcode = Arrest.Zipcode) %>%
  summarise(Arrest_Count = n())

# Calculate Average or Median Property Prices per Zip Code
property_price_stats <- zillow %>%
  group_by(Zipcode = addressZipcode) %>%
  summarise(
    Average_Price = mean(price, na.rm = TRUE),
    Median_Price = median(price, na.rm = TRUE)
  )

# Merge the Results
summary_df <- merge(arrests_per_zip, property_price_stats, by = "Zipcode")
```

```
# Save the Results
```

```
print(summary_df)
```

##	Zipcode	Arrest_Count	Average_Price	Median_Price
## 1	75104	1	329900.0	329900.0
## 2	75134	19	209900.0	209900.0
## 3	75201	3762	1994333.3	899000.0
## 4	75202	2520	297718.8	284250.0
## 5	75203	1258	329950.0	334900.0
## 6	75204	2540	594946.1	575000.0
## 7	75205	131	1024950.0	1024950.0
## 8	75206	1718	745423.0	650000.0
## 9	75208	1277	818478.5	592500.0
## 10	75209	379	3193186.8	1474499.5
## 11	75210	1184	155833.3	182500.0
## 12	75211	2369	381676.1	325000.0
## 13	75212	1228	409331.1	360000.0
## 14	75214	732	898055.7	747000.0
## 15	75215	2855	292823.3	277000.0
## 16	75216	3490	256529.4	250000.0
## 17	75217	2881	346141.4	280000.0
## 18	75218	588	718922.0	625000.0
## 19	75219	1281	413275.0	391500.0
## 20	75220	3455	1698224.8	969500.0
## 21	75223	787	529030.7	390000.0
## 22	75224	1292	431149.9	366500.0
## 23	75225	721	2524600.0	1570000.0
## 24	75226	1645	85000.0	85000.0
## 25	75227	1516	277029.6	284000.0
## 26	75228	2297	359125.9	349900.0
## 27	75229	1413	2850438.9	599000.0
## 28	75230	281	1516520.6	1099000.0
## 29	75231	2162	455094.2	440000.0
## 30	75232	1124	293549.9	287450.0
## 31	75233	333	393149.7	354999.5
## 32	75234	189	263600.0	259950.0
## 33	75235	1456	430460.0	387000.0
## 34	75236	267	332442.9	290000.0
## 35	75238	717	633026.7	624900.0
## 36	75240	1041	533422.2	450000.0
## 37	75241	1416	224970.6	239900.0
## 38	75243	2769	350712.8	309999.0
## 39	75244	145	835791.7	902400.0
## 40	75246	421	975000.0	975000.0
## 41	75248	439	959107.1	697499.5
## 42	75249	132	315527.2	349900.0
## 43	75252	231	554286.1	594700.0
## 44	75253	331	303005.8	287495.0
## 45	75254	413	683800.0	479900.0
## 46	75287	711	686076.8	530000.0

```
write.csv(summary_df, "Dallas_Arrests_Property_Summary.csv", row.names = FALSE)
```

Data visulization

```
ggplot(summary_df, aes(x = Average_Price, y = Arrest_Count)) +  
  geom_point(color = "blue") +  
  labs(  
    title = "Arrest Counts vs. Average Property Price by Zip Code",  
    x = "Average Property Price",  
    y = "Total Arrests"  
  ) +  
  theme_minimal()
```

