

# Climate-Based Sales Analysis: Fans and Cooling Devices

## Temperature Impact on Product Sales in Swiss Cities

### 1 Executive Summary

This analysis investigates the correlation between temperature and sales of cooling products (fans and air conditioners) in eight Swiss cities.

**Research question:** Is there a correlation between temperature and per-capita sales of cooling products?

### 2 1. Data & Method

We analyze sales data from Digitec Galaxus for two product categories: - **ventilator-168** (Fans) - **klimaanlage-280** (Air conditioners)

**Cities analyzed** (8): Zürich, Bern, Luzern, Basel, St. Gallen, Lugano, Lausanne, Genf

**Method:** Pearson correlation between average temperature (warmest month) and sales per 5,000 residents.

### 3 2. Data Preparation

```
# Load packages
library(tidyverse)
library(readr)
library(ggplot2)
library(knitr)
library(corrplot)
```

```
options(scipen = 999)
theme_set(theme_minimal())
```

```
# Load data
temp_data <- read_delim("Temperatur und Anzahl Personen.csv",
                        delim = ";", locale = locale(encoding = "UTF-8"))
colnames(temp_data) <- c("City", "Population", "Temperature")

sales_data <- read_csv("DigitecLive_Cleaned.csv",
                       locale = locale(encoding = "UTF-8"))
```

```
# Filter for specific cooling product categories
cooling_sales <- sales_data %>%
  filter(`infos.Category` %in% c("ventilator-168", "klimaanlage-280"))

# Filter for target cities
target_cities <- c("Zürich", "Bern", "Luzern", "Basel",
                  "St. Gallen", "Lugano", "Lausanne", "Genf")

cooling_sales_filtered <- cooling_sales %>%
  filter(cityName_clean %in% target_cities)

cat(sprintf("Total products: %d\n", nrow(cooling_sales_filtered)))
```

Total products: 278

## 4 3. Results

```
# Aggregate sales by city and calculate per-capita sales
sales_by_city <- cooling_sales_filtered %>%
  group_by(cityName_clean) %>%
  summarise(Total_Sales = n(), .groups = 'drop') %>%
  rename(City = cityName_clean) %>%
  left_join(temp_data, by = "City") %>%
  mutate(Sales_per_5k = (Total_Sales / Population) * 5000)

kable(sales_by_city %>%
      select(City, Temperature, Sales_per_5k) %>%
```

```

    arrange(desc(Temperature)),
    digits = 2,
    col.names = c("City", "Temp (°C)", "Sales per 5k"),
    caption = "Sales per 5,000 Residents by City")

```

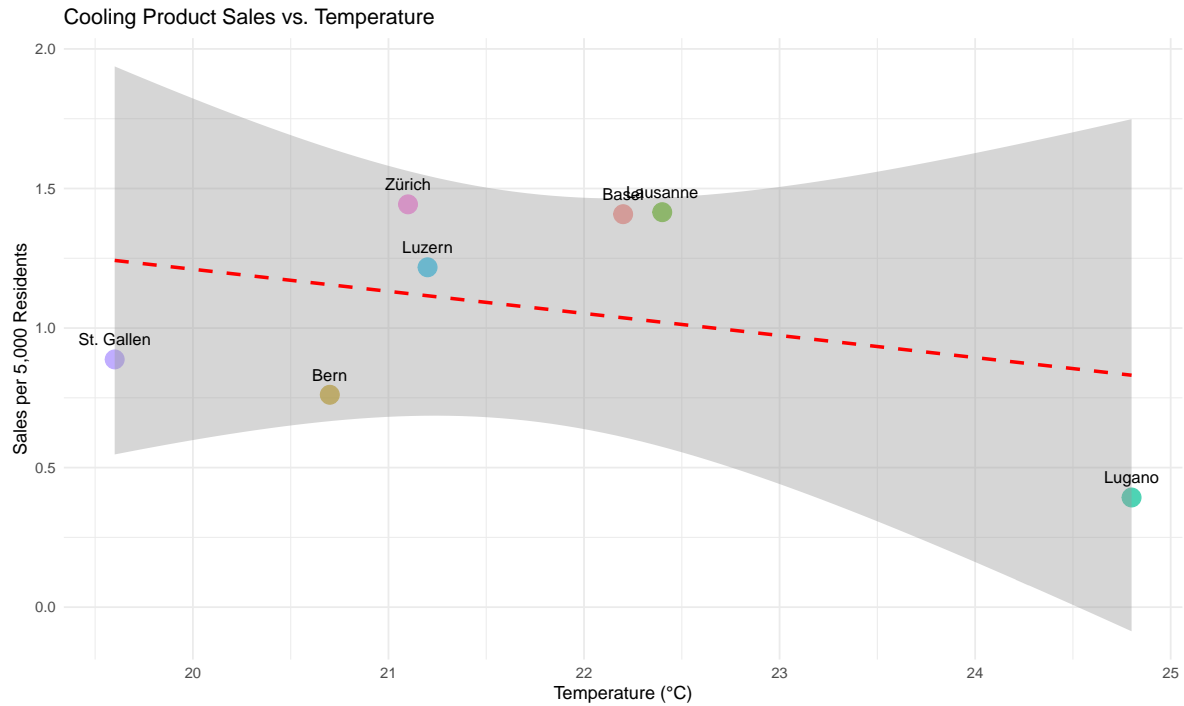
Table 1: Sales per 5,000 Residents by City

City	Temp (°C)	Sales per 5k
Lugano	24.8	0.39
Lausanne	22.4	1.42
Basel	22.2	1.41
Luzern	21.2	1.22
Zürich	21.1	1.44
Bern	20.7	0.76
St. Gallen	19.6	0.89

```

# Visualization
ggplot(sales_by_city, aes(x = Temperature, y = Sales_per_5k)) +
  geom_point(aes(color = City), size = 5, alpha = 0.7) +
  geom_smooth(method = "lm", se = TRUE, color = "red", linetype = "dashed") +
  geom_text(aes(label = City), vjust = -1.2, size = 3.5) +
  labs(
    title = "Cooling Product Sales vs. Temperature",
    x = "Temperature (°C)",
    y = "Sales per 5,000 Residents"
  ) +
  theme_minimal() +
  theme(legend.position = "none")

```



```
# Correlation analysis
cor_result <- cor.test(sales_by_city$Temperature,
                      sales_by_city$Sales_per_5k,
                      method = "pearson")

cat(sprintf("Correlation coefficient (r): %.3f\n", cor_result$estimate))
```

Correlation coefficient (r): -0.323

```
cat(sprintf("p-value: %.4f\n", cor_result$p.value))
```

p-value: 0.4798

```
cat(sprintf("Result: %s\n",
           ifelse(cor_result$p.value < 0.05,
                  "Significant correlation (p < 0.05)",
                  "No significant correlation")))
```

Result: No significant correlation

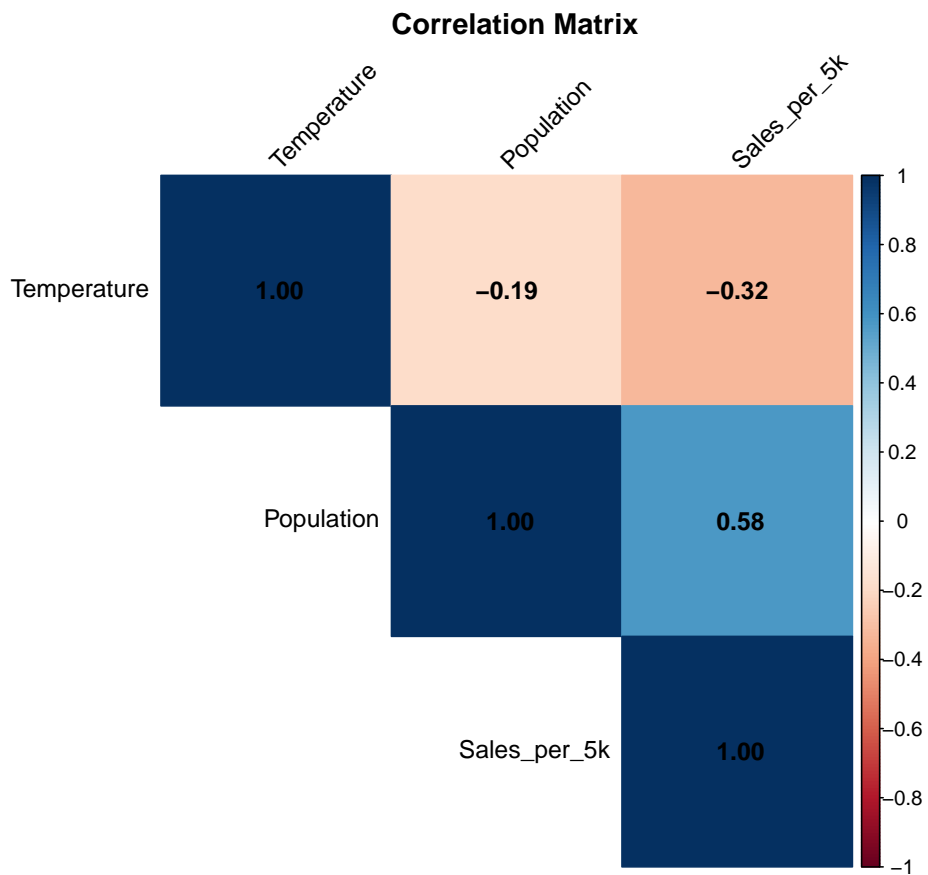
```
# Linear regression
model <- lm(Sales_per_5k ~ Temperature, data = sales_by_city)
cat(sprintf("\nR2: %.3f (%.1f%% variance explained)\n",
           summary(model)$r.squared,
           summary(model)$r.squared * 100))
```

R<sup>2</sup>: 0.104 (10.4% variance explained)

```
# Correlation matrix
cor_data <- sales_by_city %>%
  select(Temperature, Population, Sales_per_5k)

cor_matrix <- cor(cor_data, use = "complete.obs")

corrplot(cor_matrix,
          method = "color",
          type = "upper",
          addCoef.col = "black",
          tl.col = "black",
          tl.srt = 45,
          title = "Correlation Matrix",
          mar = c(0, 0, 2, 0))
```



```
# City ranking
ranking <- sales_by_city %>%
  mutate(
    Temp_Rank = rank(-Temperature),
    Sales_Rank = rank(-Sales_per_5k)
  ) %>%
  select(City, Temperature, Temp_Rank, Sales_per_5k, Sales_Rank) %>%
  arrange(Temp_Rank)

kable(ranking,
      digits = 2,
      col.names = c("City", "Temp (°C)", "Temp Rank",
                    "Sales per 5k", "Sales Rank"),
      caption = "City Ranking by Temperature and Sales")
```

Table 2: City Ranking by Temperature and Sales

City	Temp (°C)	Temp Rank	Sales per 5k	Sales Rank
Lugano	24.8	1	0.39	7
Lausanne	22.4	2	1.42	2
Basel	22.2	3	1.41	3
Luzern	21.2	4	1.22	4
Zürich	21.1	5	1.44	1
Bern	20.7	6	0.76	6
St. Gallen	19.6	7	0.89	5

## 5 4. Conclusion

**Hypothesis:** Higher temperatures correlate with higher per-capita sales of cooling products.

**Result:** NOT CONFIRMED ( $r = -0.323$ ,  $p = 0.4798$ )

- **Lugano** (24.8°C): 0.39 sales per 5k residents
- **St. Gallen** (19.6°C): 0.89 sales per 5k residents

**Interpretation:** Moderate negative correlation between temperature and sales.

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**Data:** Digitec Galaxus (ventilator-168, klimaanlage-280), BFS, MeteoSwiss