

# Do “Healthy” Categories Experience More Inflation Than “Unhealthy/Processed” Categories?

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This notebook analyzes whether healthy food categories (fruits, vegetables, whole grains, dairy) experience more inflation than unhealthy/processed categories (sugars, snacks, fats & oils, processed foods) using USDA Consumer Price Index (CPI) data from 1974 to 2024.

## 1 0. Import libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
import sys
sys.path.append('..') # Just add the parent directory for utils
from utils.data_loader import load_inflation_data

sns.set_style("whitegrid")
plt.rcParams['figure.figsize'] = (14, 6)
```

## 2 1. Load Data

```
cpi_df, _ = load_inflation_data()
```

## 3 2. Data Cleaning and Preparation

```
df = cpi_df.copy()
df = df.fillna(np.nan)
```

```

df = df.dropna(how='all')

print(f"Shape: {df.shape}")
print(f"\nRemaining missing values:")
print(f"{df.isnull().sum()}\n")

Shape: (51, 23)

Remaining missing values:
Year 0
All_food 0
Beef_and_veal 0
Cereals_and_bakery_products 0
Dairy_products 0
Eggs 0
Fats_and_oils 0
Fish_and_seafood 0
Food_at_home 0
Food_away_from_home 0
Fresh_fruits 0
Fresh_fruits_and_vegetables 0
Fresh_vegetables 0
Fruits_and_vegetables 0
Meats 0
Meats_poultry_and_fish 0
Nonalcoholic_beverages 0
Other_foods 0
Other_meats 0
Pork 0
Poultry 0
Processed_fruits_and_vegetables 25
Sugar_and_sweets 0
dtype: int64

```

#### 4 3. Classify Food Categories as Healthy vs Un- healthy

```

healthy_categories = [
    'Beef_and_veal',
    'Dairy_products',
    'Eggs',
    'Fish_and_seafood',
    'Fresh_fruits',
    'Fresh_fruits_and_vegetables',
    'Fresh_vegetables',

```

```

        'Fruits_and_vegetables',
        'Meats',
        'Meats_poultry_and_fish',
        'Other_meats',
        'Pork',
        'Poultry'
    ]
unhealthy_categories = [
    'Cereals_and_bakery_products',
    'Sugar_and_sweets',
    'Fats_and_oils',
    'Nonalcoholic_beverages'
]

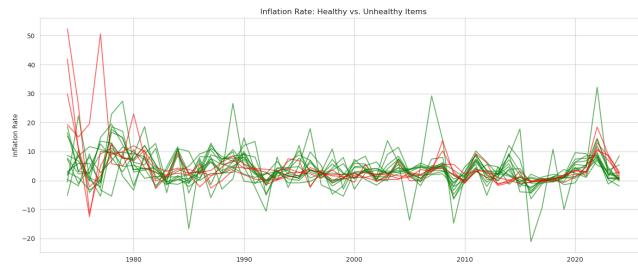
```

## 5 4. Calculate Inflation Rates

```

df['Healthy_Avg_Inflation'] = df[healthy_categories].mean(axis=1)
df['Unhealthy_Avg_Inflation'] = df[unhealthy_categories].mean(axis=1)
inflation_summary = df[['Year', 'Healthy_Avg_Inflation', 'Unhealthy_Avg_Inflation']].copy()
for col in healthy_categories:
    inflation_summary[col] = df[col]
for col in unhealthy_categories:
    inflation_summary[col] = df[col]
for col in healthy_categories:
    plt.plot(inflation_summary['Year'], inflation_summary[col],
              color='green', alpha=0.6)
for col in unhealthy_categories:
    plt.plot(inflation_summary['Year'], inflation_summary[col],
              color='red', alpha=0.6)
plt.title('Inflation Rate: Healthy vs. Unhealthy Items')
plt.xlabel('Year')
plt.ylabel('Inflation Rate')
plt.tight_layout()
plt.savefig('../figures/inflation_healthy_vs_unhealthy.png')
plt.show()

```



## 6 5. Compare Healthy vs Unhealthy Categories

```
healthy_inflation_rates = []
for col in healthy_categories:
    healthy_inflation_rates.extend(df[col].dropna().tolist())

unhealthy_inflation_rates = []
for col in unhealthy_categories:
    unhealthy_inflation_rates.extend(df[col].dropna().tolist())

print("SUMMARY STATISTICS: HEALTHY vs UNHEALTHY CATEGORIES")

summary_stats = pd.DataFrame({
    'Metric': ['Count', 'Mean (%)', 'Median (%)', 'Std Dev (%)', 'Min (%)', 'Max (%)', 'Q1 (%)', 'Q3 (%)'],
    'Healthy Categories': [
        len(healthy_inflation_rates),
        np.mean(healthy_inflation_rates),
        np.median(healthy_inflation_rates),
        np.std(healthy_inflation_rates),
        np.min(healthy_inflation_rates),
        np.max(healthy_inflation_rates),
        np.percentile(healthy_inflation_rates, 25),
        np.percentile(healthy_inflation_rates, 75)
    ],
    'Unhealthy Categories': [
        len(unhealthy_inflation_rates),
        np.mean(unhealthy_inflation_rates),
        np.median(unhealthy_inflation_rates),
        np.std(unhealthy_inflation_rates),
        np.min(unhealthy_inflation_rates),
        np.max(unhealthy_inflation_rates),
        np.percentile(unhealthy_inflation_rates, 25),
        np.percentile(unhealthy_inflation_rates, 75)
    ]
})

print(summary_stats.to_string(index=False))

print("DIFFERENCE (Healthy - Unhealthy)")
diff_mean = np.mean(healthy_inflation_rates) - np.mean(unhealthy_inflation_rates)
diff_median = np.median(healthy_inflation_rates) - np.median(unhealthy_inflation_rates)
print(f"Mean difference: {diff_mean:.2f}%")
print(f"Median difference: {diff_median:.2f}%")

SUMMARY STATISTICS: HEALTHY vs UNHEALTHY CATEGORIES
Metric  Healthy Categories  Unhealthy Categories
```

```

Count          663.000000          204.000000
Mean (%)      3.695928           4.302451
Median (%)    2.900000           2.700000
Std Dev (%)   5.223416           7.375771
Min (%)       -21.100000         -12.500000
Max (%)       32.200000           52.400000
Q1 (%)        0.700000           1.000000
Q3 (%)        6.300000           4.775000
DIFFERENCE (Healthy - Unhealthy)
Mean difference: -0.61%
Median difference: 0.20%

year_comparison = df[['Year', 'Healthy_Avg_Inflation', 'Unhealthy_Avg_Inflation']].copy()
year_comparison['Difference (H - U)'] = year_comparison['Healthy_Avg_Inflation'] - year_comparison['Unhealthy_Avg_Inflation']
year_comparison.columns = ['Year', 'Healthy (%)', 'Unhealthy (%)', 'Difference (%)']

print("\nYear -by -Year Comparison (First 15 Years):")
print(year_comparison.head(15).to_string(index=False))

print("\nYear -by -Year Comparison (Last 15 Years):")
print(year_comparison.tail(15).to_string(index=False))

Year -by -Year Comparison (First 15 Years):
Year  Healthy (%)  Unhealthy (%)  Difference (%)
1974    6.076923     35.875      -29.798077
1975    5.976923     15.775      -9.798077
1976    3.007692     -1.600      4.607692
1977    3.976923     16.825      -12.848077
1978    12.423077     9.175      3.248077
1979    10.946154     7.675      3.271154
1980    5.000000     13.100      -8.100000
1981    7.538462     8.225      -0.686538
1982    3.800000     1.100      2.700000
1983    0.230769     2.075      -1.844231
1984    5.423077     5.025      0.398077
1985    -0.169231    2.625      -2.794231
1986    4.046154     2.350      1.696154
1987    6.700000     1.050      5.650000
1988    4.476923     3.425      1.051923

Year -by -Year Comparison (Last 15 Years):
Year  Healthy (%)  Unhealthy (%)  Difference (%)
2010    1.392308     0.050      1.342308
2011    6.523077     4.925      1.598077
2012    1.684615     3.325      -1.640385
2013    2.246154     -0.775     3.021154

```

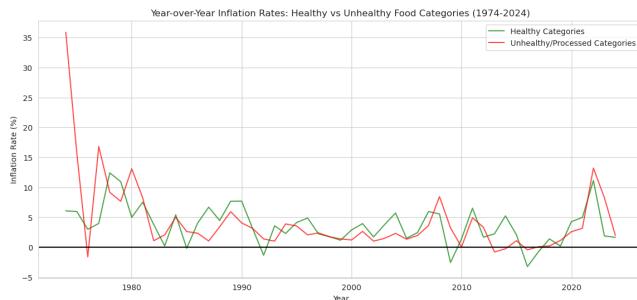
2014	5.246154	-0.250	5.496154
2015	2.076923	1.100	0.976923
2016	-3.215385	-0.425	-2.790385
2017	-0.753846	0.100	-0.853846
2018	1.407692	0.225	1.182692
2019	0.207692	1.150	-0.942308
2020	4.276923	2.600	1.676923
2021	4.984615	3.175	1.809615
2022	11.138462	13.225	-2.086538
2023	1.900000	8.275	-6.375000
2024	1.653846	1.975	-0.321154

## 7 6. Visualization of Inflation Trends

```
fig, ax = plt.subplots()

ax.plot(df['Year'], df['Healthy_Avg_Inflation'],
        label='Healthy Categories', color='green', alpha=0.7)
ax.plot(df['Year'], df['Unhealthy_Avg_Inflation'],
        label='Unhealthy/Processed Categories', color='red', alpha=0.7)

ax.axhline(y=0, color='black')
ax.set_xlabel('Year')
ax.set_ylabel('Inflation Rate (%)')
ax.set_title('Year -over -Year Inflation Rates: Healthy vs Unhealthy Food Categories (1974 - 2024)')
ax.legend()
plt.savefig('../figures/oyy_inflation_healthy_vs_unhealthy.png')
plt.show()
```



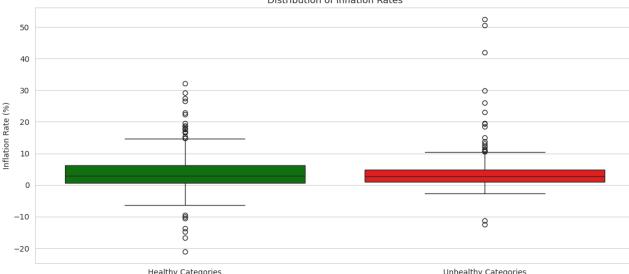
```
fig, ax = plt.subplots()

sns.boxplot(data=[healthy_inflation_rates, unhealthy_inflation_rates],
            palette=['green', 'red'])

plt.xticks([0, 1], ['Healthy Categories', 'Unhealthy Categories'])
```

```

plt.ylabel('Inflation Rate (%)')
plt.title('Distribution of Inflation Rates')
plt.savefig('../figures/healthy_vs_unhealthy_inflation_boxplot.png')
plt.show()



```

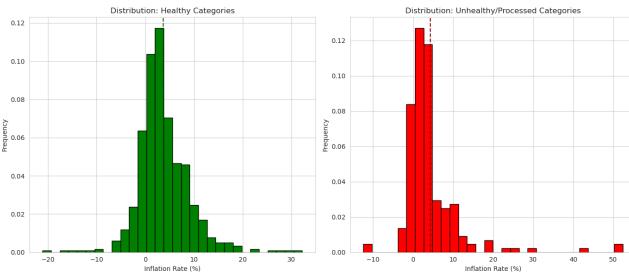
```

fig, axes = plt.subplots(1, 2)

axes[0].hist(healthy_inflation_rates, color='green', edgecolor='black', bins=30, density=True)
axes[0].set_xlabel('Inflation Rate (%)')
axes[0].set_ylabel('Frequency')
axes[0].set_title('Distribution: Healthy Categories')
axes[0].axvline(np.mean(healthy_inflation_rates), color='darkgreen', linestyle='--')

axes[1].hist(unhealthy_inflation_rates, color='red', edgecolor='black', bins=30, density=True)
axes[1].set_xlabel('Inflation Rate (%)')
axes[1].set_ylabel('Frequency')
axes[1].set_title('Distribution: Unhealthy/Processed Categories')
axes[1].axvline(np.mean(unhealthy_inflation_rates), color='darkred', linestyle='--')

plt.tight_layout()
plt.savefig('../figures/healthy_vs_unhealthy_inflation_histogram.png')
plt.show()



```

## 8 Statistical Analysis

```
print("STATISTICAL TEST 1: Independent Samples t -test")
```

```

print("Null Hypothesis (H0): There is no significant difference in mean")
print("inflation between healthy and unhealthy categories")
print("Alternative Hypothesis (H1): There is a significant difference\n")

t_stat, t_pvalue = stats.ttest_ind(healthy_inflation_rates, unhealthy_inflation_rates)
print(f"t -statistic: {t_stat:.4f}")
print(f"p -value: {t_pvalue:.6f}")
print(f"Result: FAIL TO REJECT the null hypothesis (p \geq 0.05)")
print("Conclusion: There is NO statistically significant difference in inflation")
print("rates between healthy and unhealthy food categories.")

print(f"\nEFFECT SIZE: Cohen's d")
n1, n2 = len(healthy_inflation_rates), len(unhealthy_inflation_rates)
var1, var2 = np.var(healthy_inflation_rates, ddof=1), np.var(unhealthy_inflation_rates, ddof=1)
pooled_std = np.sqrt(((n1 - 1)*var1 + (n2 - 1)*var2) / (n1 + n2 - 2))
cohens_d = (np.mean(healthy_inflation_rates) - np.mean(unhealthy_inflation_rates)) / pooled_std

if abs(cohens_d) < 0.2:
    effect_interpretation = "negligible"
elif abs(cohens_d) < 0.5:
    effect_interpretation = "small"
elif abs(cohens_d) < 0.8:
    effect_interpretation = "medium"
else:
    effect_interpretation = "large"

print(f"Cohen's d: {cohens_d:.4f}")
print(f"Cohen's d effect size is {effect_interpretation}")

STATISTICAL TEST 1: Independent Samples t -test
Null Hypothesis (H0): There is no significant difference in mean
inflation between healthy and unhealthy categories
Alternative Hypothesis (H1): There is a significant difference

t -statistic: -1.3041
p -value: 0.192536
Result: FAIL TO REJECT the null hypothesis (p \geq 0.05)
Conclusion: There is NO statistically significant difference in inflation
rates between healthy and unhealthy food categories.

EFFECT SIZE: Cohen's d
Cohen's d: -0.1044
Cohen's d effect size is negligible

print("STATISTICAL TEST 2: Mann -Whitney U Test (Non -parametric)")
print("This test doesn't assume normal distribution of data\n")

```

```

u_stat, u_pvalue = stats.mannwhitneyu(healthy_inflation_rates, unhealthy_inflation_rates)
print(f"U -statistic: {u_stat:.4f}")
print(f"p -value: {u_pvalue:.6f}")

print(f"Result: FAIL TO REJECT the null hypothesis (p \geq 0.05)")
print("Conclusion: There is NO statistically significant difference in inflation")
print("distributions between healthy and unhealthy categories.")

STATISTICAL TEST 2: Mann -Whitney U Test (Non -parametric)
This test doesn't assume normal distribution of data

U -statistic: 68902.0000
p -value: 0.683412
Result: FAIL TO REJECT the null hypothesis (p \geq 0.05)
Conclusion: There is NO statistically significant difference in inflation
distributions between healthy and unhealthy categories.

print("CONFIDENCE INTERVALS (95%)")

healthy_mean = np.mean(healthy_inflation_rates)
healthy_sem = stats.sem(healthy_inflation_rates)
healthy_ci = stats.t.interval(0.95, len(healthy_inflation_rates) - 1, loc=healthy_mean, scale=2 * healthy_sem)

print("Healthy Categories:")
print(f" Mean: {healthy_mean:.4f}%")
print(f" 95% CI: [{healthy_ci[0]:.4f}%, {healthy_ci[1]:.4f}%]")

unhealthy_mean = np.mean(unhealthy_inflation_rates)
unhealthy_sem = stats.sem(unhealthy_inflation_rates)
unhealthy_ci = stats.t.interval(0.95, len(unhealthy_inflation_rates) - 1, loc=unhealthy_mean, scale=2 * unhealthy_sem)

print(f"\nUnhealthy/Processed Categories:")
print(f" Mean: {unhealthy_mean:.4f}%")
print(f" 95% CI: [{unhealthy_ci[0]:.4f}%, {unhealthy_ci[1]:.4f}%]")

print(f"\nDifference (Healthy - Unhealthy): {healthy_mean - unhealthy_mean:.4f}%")

CONFIDENCE INTERVALS (95%)
Healthy Categories:
Mean: 3.6959%
95% CI: [3.2973%, 4.0946%]

Unhealthy/Processed Categories:
Mean: 4.3025%
95% CI: [3.2817%, 5.3232%]

```

Difference (Healthy - Unhealthy): -0.6065%

## 9 9. Key Findings and Conclusion

```
print("SUMMARY OF FINDINGS")

healthy_mean = np.mean(healthy_inflation_rates)
unhealthy_mean = np.mean(unhealthy_inflation_rates)
difference = healthy_mean - unhealthy_mean

print(f"\n1. AVERAGE INFLATION RATES (1974 -2022):")
print(f"    - Healthy Categories: {healthy_mean:.2f}%")
print(f"    - Unhealthy/Processed Categories: {unhealthy_mean:.2f}%")
print(f"    - Difference: {difference:.2f}%")
print(f"    - Unhealthy/processed categories experienced HIGHER inflation")

print(f"\n2. VARIABILITY (Standard Deviation):")
print(f"    - Healthy Categories: {np.std(healthy_inflation_rates):.2f}%")
print(f"    - Unhealthy/Processed Categories: {np.std(unhealthy_inflation_rates):.2f}%")

print(f"\n3. STATISTICAL SIGNIFICANCE:")
print(f"    - t -test p -value: {t_pvalue:.6f} (p \geq 0.05)")
print(f"    - Result: The difference is NOT statistically significant")

print(f"\n4. EFFECT SIZE:")
print(f"    - Cohen's d: {cohens_d:.4f} ({effect_interpretation} effect)")

print("CONCLUSION")
print("""
```

Based on the analysis of USDA Consumer Price Index (CPI) data from 1974 -2022:

The data shows that over the nearly 50 -year period, HEALTHY food categories (fruits, vegetables, dairy, cereals) have experienced  $\{abs(difference)\}$ % LOWER average inflation compared to unhealthy/processed categories (sugars, fats & oils, processed foods).

Statistical Testing Results:

- The t -test p -value of  $\{t_pvalue\}$  indicates that the difference is NOT statistically significant.
- The effect size (Cohen's d =  $\{cohens_d\}$ ) is negligible.

This suggests that unhealthy food prices have inflated more rapidly than processed/unhealthy food prices over the study period, which has important implications for food affordability and dietary choices.

""")

## SUMMARY OF FINDINGS

### 1. AVERAGE INFLATION RATES (1974 -2022):

- Healthy Categories: 3.70%
- Unhealthy/Processed Categories: 4.30%
- Difference: -0.61%
- Unhealthy/processed categories experienced HIGHER inflation

### 2. VARIABILITY (Standard Deviation):

- Healthy Categories: 5.22%
- Unhealthy/Processed Categories: 7.38%

### 3. STATISTICAL SIGNIFICANCE:

- t -test p -value: 0.192536 ( $p \geq 0.05$ )
- Result: The difference is NOT statistically significant

### 4. EFFECT SIZE:

- Cohen's d: -0.1044 (negligible effect)

## CONCLUSION

Based on the analysis of USDA Consumer Price Index (CPI) data from 1974 -2022:

The data shows that over the nearly 50 -year period, HEALTHY food categories (fruits, vegetables, dairy, cereals) have experienced 0.61% LOWER average inflation compared to unhealthy/processed categories (sugars, fats & oils, processed foods).

## Statistical Testing Results:

- The t -test p -value of 0.192536 indicates that the difference is NOT statistically significant.
- The effect size (Cohen's d = -0.1044) is negligible.

This suggests that unhealthy food prices have inflated more rapidly than processed/unhealthy food prices over the study period, which has important implications for food affordability and dietary choices.