

```
In [89]: !pip install numpy pandas matplotlib seaborn
```

```
Requirement already satisfied: numpy in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (2.2.5)
Requirement already satisfied: pandas in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (2.2.3)
Requirement already satisfied: matplotlib in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (3.10.1)
Collecting seaborn
  Downloading seaborn-0.13.2-py3-none-any.whl.metadata (5.4 kB)
Requirement already satisfied: python-dateutil>=2.8.2 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from pandas) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from pandas) (2025.2)
Requirement already satisfied: contourpy>=1.0.1 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (1.3.2)
Requirement already satisfied: cycler>=0.10 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (4.57.0)
Requirement already satisfied: kiwisolver>=1.3.1 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (1.4.8)
Requirement already satisfied: packaging>=20.0 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (25.0)
Requirement already satisfied: pillow>=8 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (11.2.1)
Requirement already satisfied: pyparsing>=2.3.1 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from matplotlib) (3.2.3)
Requirement already satisfied: six>=1.5 in /Users/yaswanthkumarvejandla/yash/lib/python3.12/site-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
Downloading seaborn-0.13.2-py3-none-any.whl (294 kB)
Installing collected packages: seaborn
Successfully installed seaborn-0.13.2
```

```
In [91]: import pandas as pd
```

```
In [93]: data = pd.read_csv("daily_sales_23_25.csv")
```

```
In [97]: data.columns = [
    'period_no',
    'week_no',
    'date',
    'store_no',
    'store_name',
    'total_revenue',
    'projected_psa',
    'last_week_revenue',
    'last_week_pct_change',
    'last_year_revenue',
    'last_year_pct_change',
    'lunch_sales',
```

```
'lunch_pct',  
'labor_pct',  
'scheduled_pct',  
'labor_cost',  
'scheduled_hours',  
'actual_hours',  
'hours_difference',  
'dpmh',  
'customer_count',  
'avg_ticket',  
'status'  
]
```

```
In [101... # Convert date column  
data['date'] = pd.to_datetime(data['date'], errors='coerce')  
  
# Convert numeric columns  
numeric_cols = [  
    'total_revenue', 'projected_psa', 'last_week_revenue',  
    'last_week_pct_change', 'last_year_revenue', 'last_year_pct_change',  
    'lunch_sales', 'lunch_pct', 'labor_pct', 'scheduled_pct', 'labor_cost',  
    'scheduled_hours', 'actual_hours', 'hours_difference',  
    'dpmh', 'customer_count', 'avg_ticket'  
]  
  
data[numeric_cols] = data[numeric_cols].apply(pd.to_numeric, errors='coerce')
```

```
In [103... # Show count of missing values  
print(data.isna().sum())  
  
# Drop rows with missing dates or revenue  
data = data.dropna(subset=['date', 'total_revenue'])  
  
# Optionally, fill remaining missing numeric values with 0 or mean  
data[numeric_cols] = data[numeric_cols].fillna(0)
```

```

period_no      0
week_no        0
date           0
store_no       0
store_name     0
total_revenue  0
projected_psa  0
last_week_revenue  0
last_week_pct_change  0
last_year_revenue  0
last_year_pct_change  0
lunch_sales    0
lunch_pct      0
labor_pct      0
scheduled_pct  0
labor_cost     0
scheduled_hours  0
actual_hours   0
hours_difference  0
dpmh           0
customer_count 0
avg_ticket     0
status         0
dtype: int64

```

```
In [105... data = data.sort_values(by='date')
```

```
In [107... import numpy as np
```

```

sales = data['total_revenue'].to_numpy()
dates = data['date'].to_numpy()

print("Total revenue: $", np.sum(sales))
print("Average daily sales: $", np.mean(sales))
print("Max sale: $", sales.max(), "on", dates[np.argmax(sales)])
print("Min sale: $", sales.min(), "on", dates[np.argmin(sales)])

```

Total revenue: \$ 1433697.25

Average daily sales: \$ 2222.786434108527

Max sale: \$ 4108.74 on 2024-03-29T00:00:00.000000000

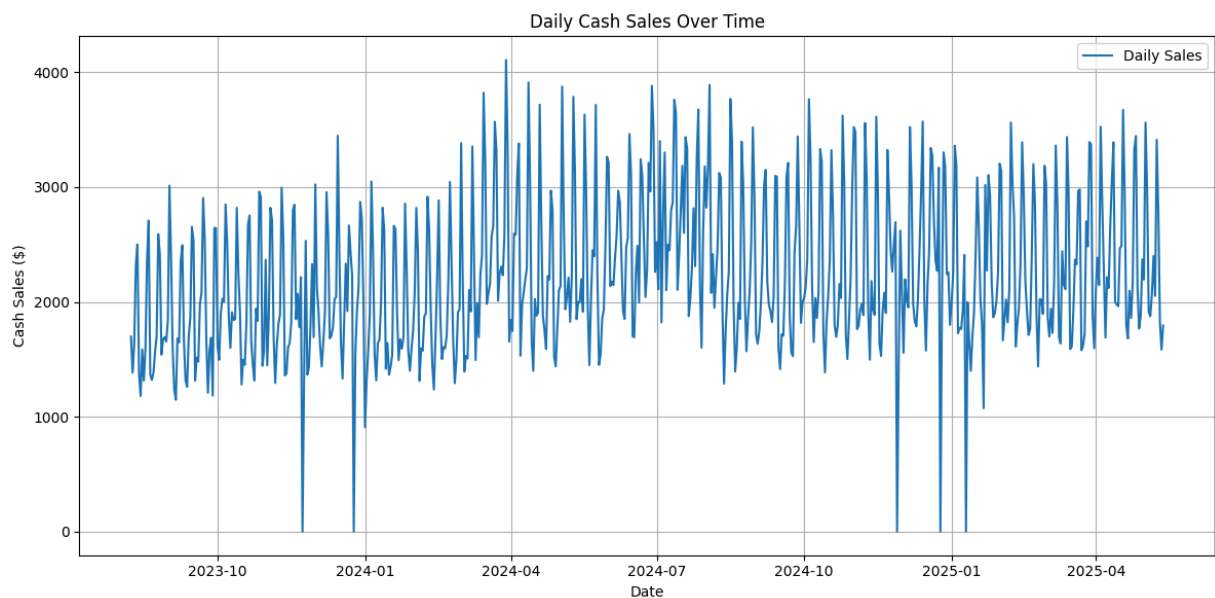
Min sale: \$ 0.0 on 2023-11-23T00:00:00.000000000

```
In [109... import matplotlib.pyplot as plt
```

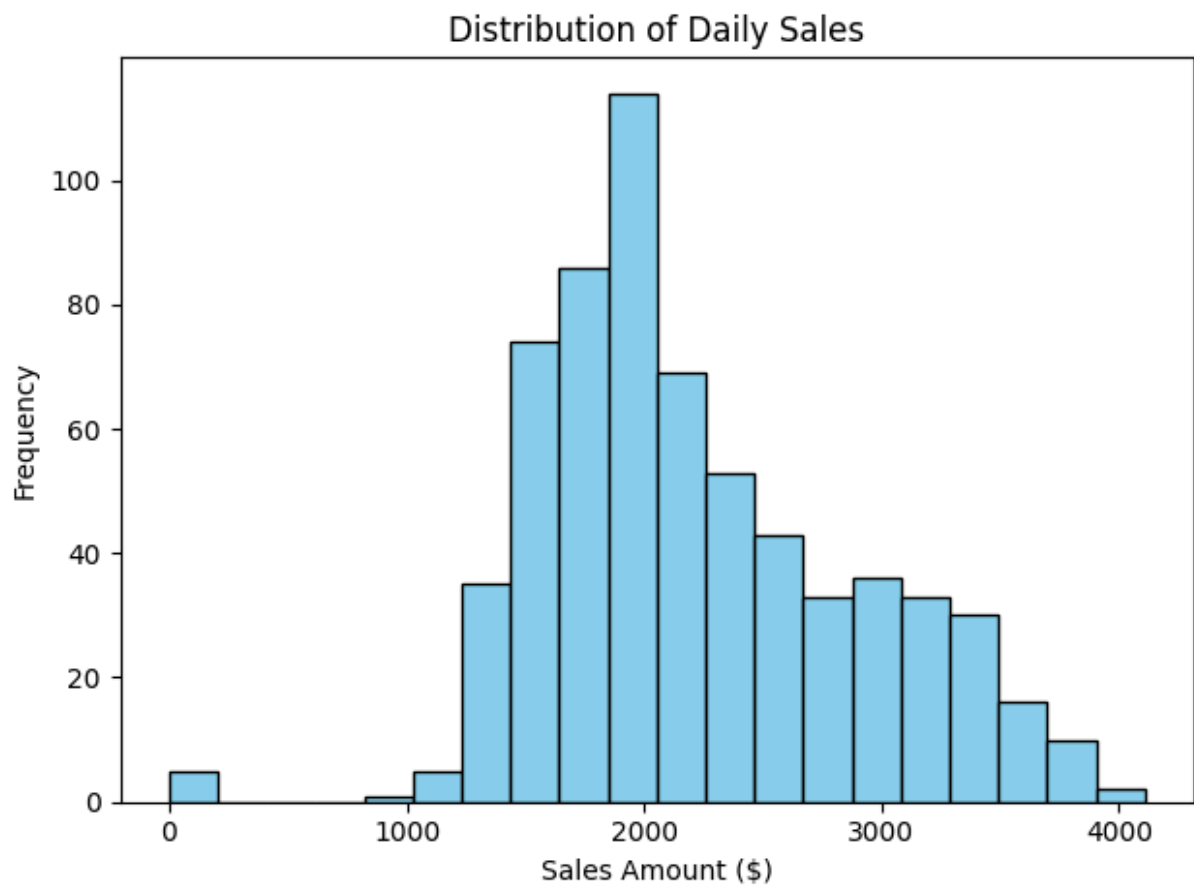
```

plt.figure(figsize=(12, 6))
plt.plot(data['date'], data['total_revenue'], label='Daily Sales')
plt.title("Daily Cash Sales Over Time")
plt.xlabel("Date")
plt.ylabel("Cash Sales ($)")
plt.grid(True)
plt.legend()
plt.tight_layout()
plt.show()

```



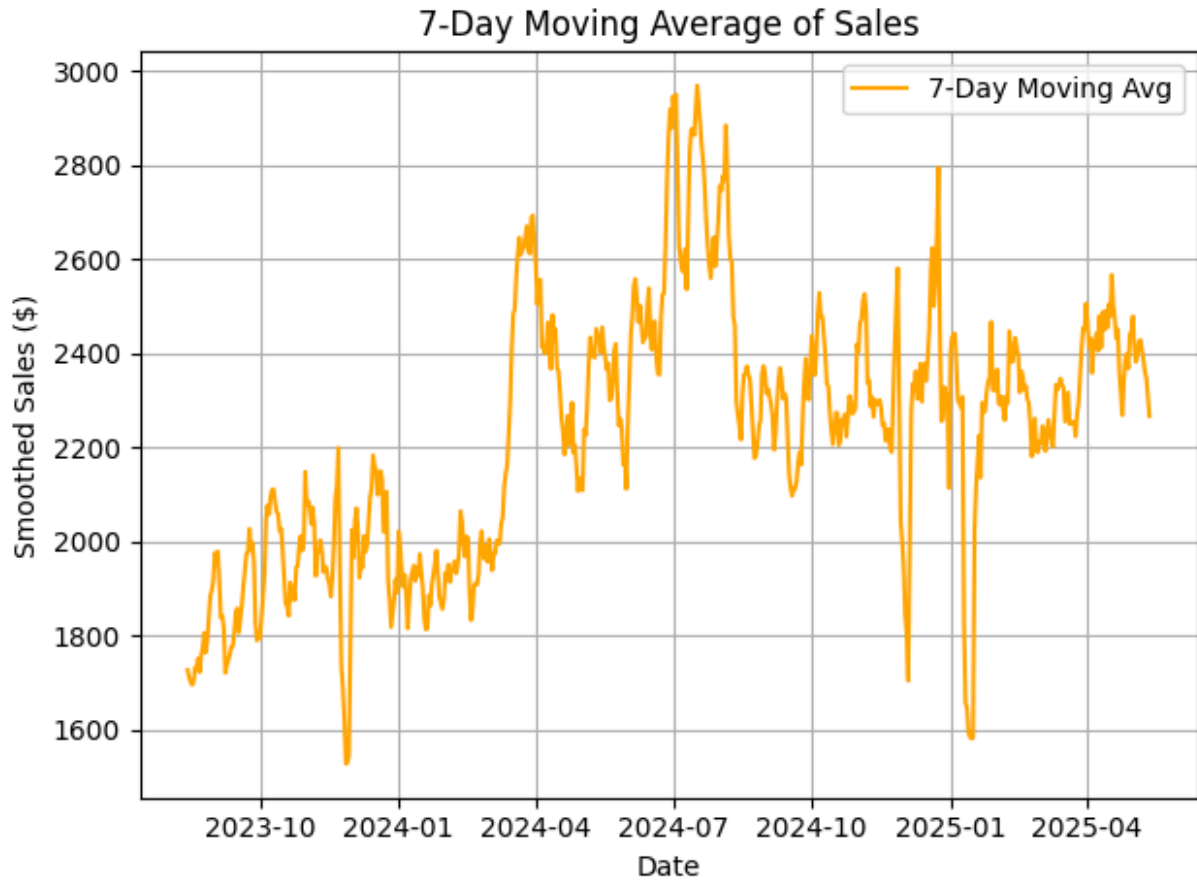
```
In [111]: plt.hist(sales, bins=20, color='skyblue', edgecolor='black')
plt.title("Distribution of Daily Sales")
plt.xlabel("Sales Amount ($)")
plt.ylabel("Frequency")
plt.tight_layout()
plt.show()
```



```
In [113]: moving_avg = np.convolve(sales, np.ones(7)/7, mode='valid')

plt.plot(data['date'][6:], moving_avg, color='orange', label='7-Day Moving A
```

```
plt.title("7-Day Moving Average of Sales")
plt.xlabel("Date")
plt.ylabel("Smoothed Sales ($)")
plt.grid(True)
plt.legend()
plt.tight_layout()
plt.show()
```



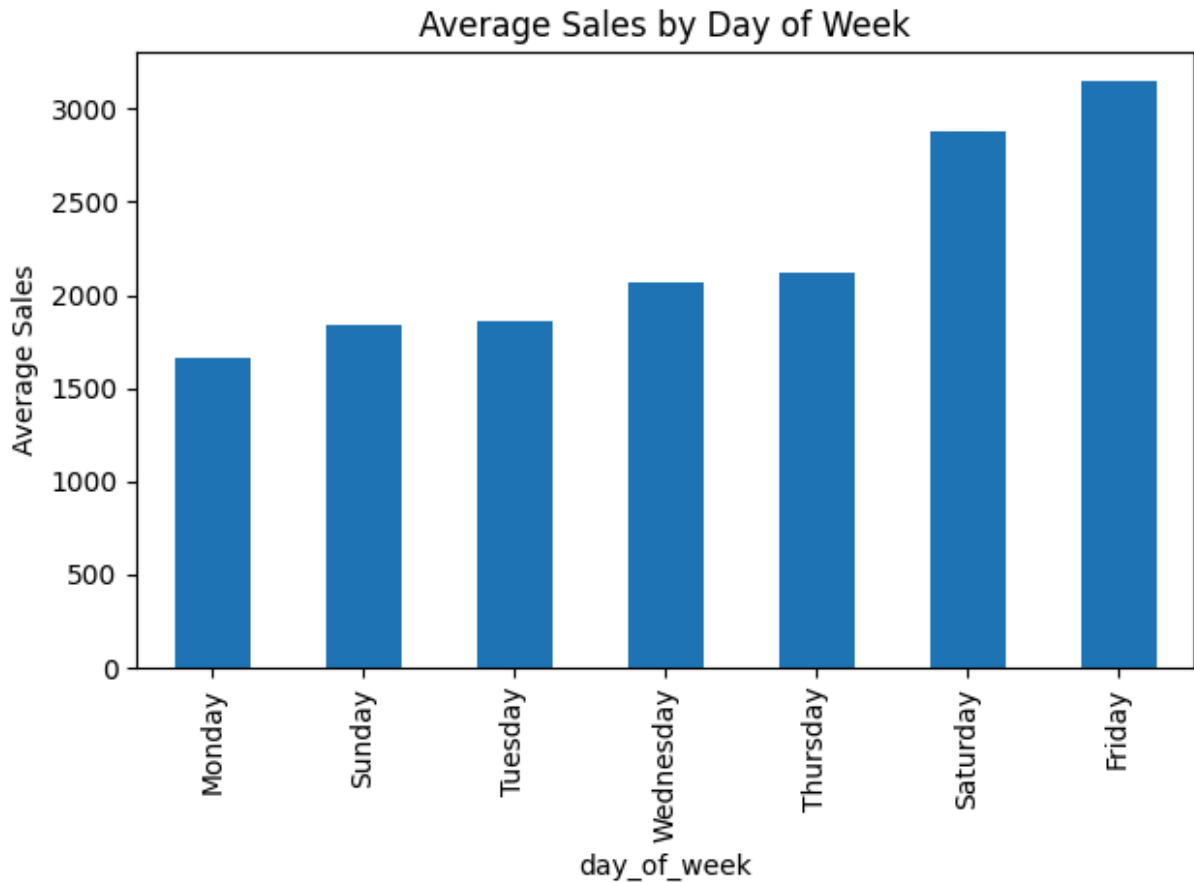
```
In [115... top_days = data.sort_values(by='total_revenue', ascending=False).head(10)
print(top_days[['date', 'total_revenue']])
```

	date	total_revenue
410	2024-03-29	4108.74
396	2024-04-12	3911.43
283	2024-08-03	3890.77
319	2024-06-28	3882.43
375	2024-05-03	3875.77
424	2024-03-15	3822.61
368	2024-05-10	3787.33
270	2024-08-16	3768.02
221	2024-10-04	3766.21
305	2024-07-12	3761.20

```
In [117... data['day_of_week'] = data['date'].dt.day_name()
avg_by_day = data.groupby('day_of_week')['total_revenue'].mean().sort_values

avg_by_day.plot(kind='bar', title="Average Sales by Day of Week")
plt.ylabel("Average Sales")
```

```
plt.tight_layout()
plt.show()
```



```
In [121... import matplotlib.pyplot as plt
import seaborn as sns

# Grouped data
data['day_of_week'] = data['date'].dt.day_name()
data['week'] = data['date'].dt.isocalendar().week
data['month'] = data['date'].dt.to_period('M')
data['labor_per_revenue'] = data['labor_cost'] / data['total_revenue']
```

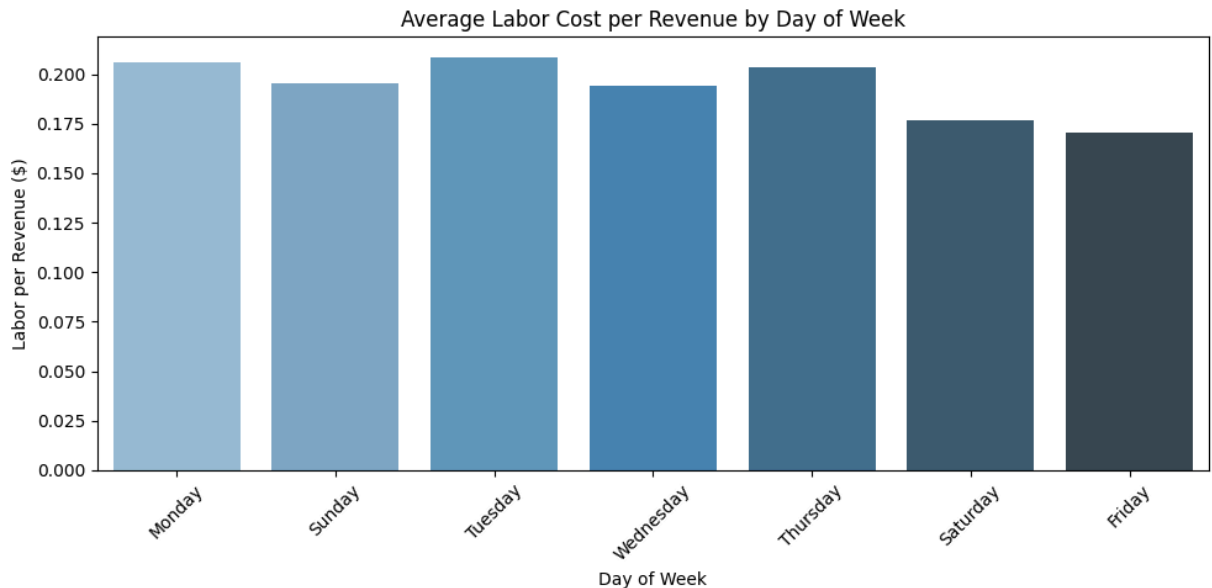
```
In [123... day_avg = data.groupby('day_of_week')[['total_revenue', 'labor_cost', 'labor_per_revenue']]
week_avg = data.groupby('week')[['total_revenue', 'labor_cost', 'labor_per_revenue']]
month_avg = data.groupby('month')[['total_revenue', 'labor_cost', 'labor_per_revenue']]
```

```
In [125... plt.figure(figsize=(10, 5))
sns.barplot(x=day_avg.index, y=day_avg['labor_per_revenue'], palette='Blues')
plt.title("Average Labor Cost per Revenue by Day of Week")
plt.ylabel("Labor per Revenue ($)")
plt.xlabel("Day of Week")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

```
/var/folders/md/yz4wns293812bb6m3pw1zpj0000gn/T/ipykernel_17504/392813891.p
y:2: FutureWarning:
```

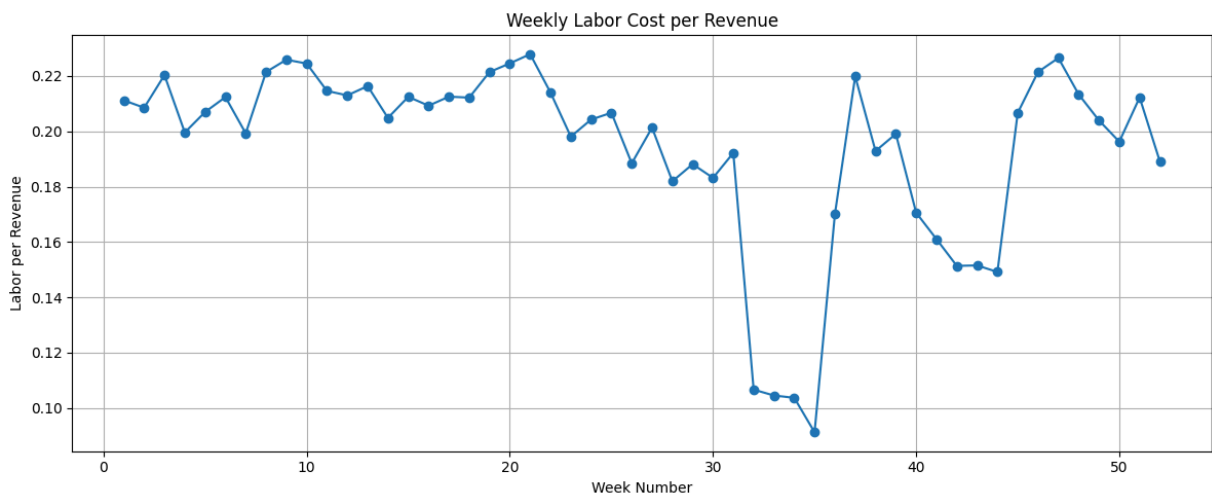
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=day_avg.index, y=day_avg['labor_per_revenue'], palette='Blue
s_d')
```



In [127...

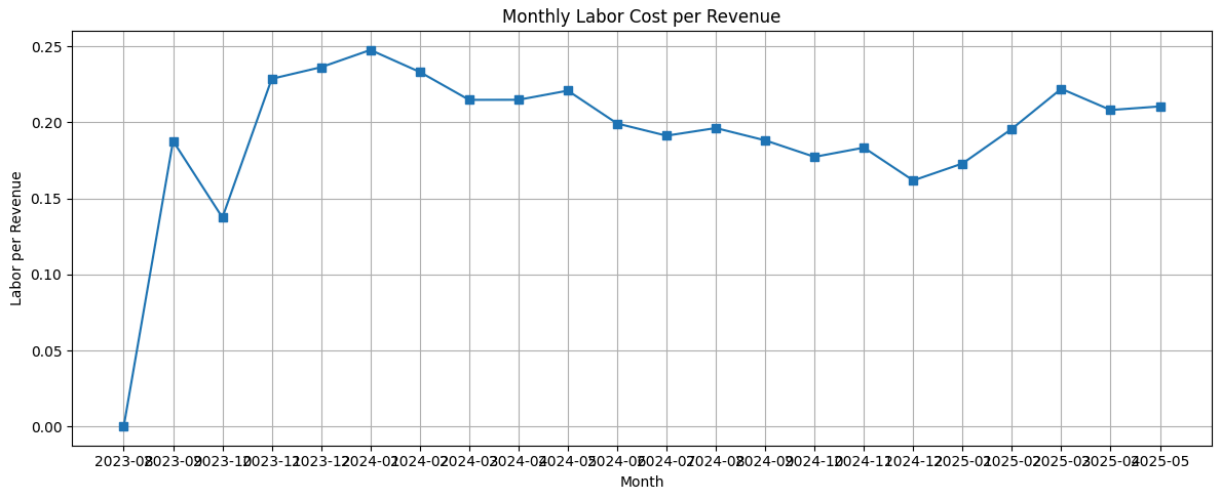
```
plt.figure(figsize=(12, 5))
plt.plot(week_avg.index, week_avg['labor_per_revenue'], marker='o')
plt.title("Weekly Labor Cost per Revenue")
plt.xlabel("Week Number")
plt.ylabel("Labor per Revenue")
plt.grid(True)
plt.tight_layout()
plt.show()
```



In [129...

```
plt.figure(figsize=(12, 5))
plt.plot(month_avg.index.astype(str), month_avg['labor_per_revenue'], marker
plt.title("Monthly Labor Cost per Revenue")
```

```
plt.xlabel("Month")
plt.ylabel("Labor per Revenue")
plt.grid(True)
plt.tight_layout()
plt.show()
```



In [135... **import** pandas **as** pd

```
# Make sure 'efficiency_ratio' exists
data['efficiency_ratio'] = data['total_revenue'] / data['labor_cost']

# Create a day of week column
data['day_of_week'] = data['date'].dt.day_name()

# Group by day and calculate average efficiency
day_efficiency = data.groupby('day_of_week')['efficiency_ratio'].mean().reset_index()

# Classify each day
day_efficiency['category'] = day_efficiency['efficiency_ratio'].apply(
    lambda x: 'More Employees than Revenue' if x < 1 else 'More Revenue than Employees'
)

# Sort for consistent weekday order
from pandas.api.types import CategoricalDtype
weekday_order = ['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Sunday']
day_efficiency['day_of_week'] = day_efficiency['day_of_week'].astype(CategoricalDtype(weekday_order))
day_efficiency = day_efficiency.sort_values('day_of_week')

print(day_efficiency)
```

	day_of_week	efficiency_ratio	category
1	Monday	inf	More Revenue than Employees
5	Tuesday	inf	More Revenue than Employees
6	Wednesday	inf	More Revenue than Employees
4	Thursday	inf	More Revenue than Employees
0	Friday	inf	More Revenue than Employees
2	Saturday	inf	More Revenue than Employees
3	Sunday	inf	More Revenue than Employees

In [139... **import** pandas **as** pd


```

# Ensure date column is in datetime format
data['date'] = pd.to_datetime(data['date'], errors='coerce')

# Extract year and month
data['year'] = data['date'].dt.year
data['month'] = data['date'].dt.month

# Group by year and month, summing revenue
monthly_revenue = data.groupby(['year', 'month'])['total_revenue'].sum().reset_index()

# Pivot the table so each year is a column, months are rows
monthly_pivot = monthly_revenue.pivot(index='month', columns='year', values='total_revenue')

# Compute year-over-year growth %
monthly_pivot['growth_23_to_24'] = ((monthly_pivot[2024] - monthly_pivot[2023]) / monthly_pivot[2023]) * 100
monthly_pivot['growth_24_to_25'] = ((monthly_pivot[2025] - monthly_pivot[2024]) / monthly_pivot[2024]) * 100

# Round values for presentation
monthly_pivot = monthly_pivot.round(2)

# Display the result
print(monthly_pivot)

```

	year	2023	2024	2025	growth_23_to_24	growth_24_to_25
month						
1		NaN	57295.18	66827.29	NaN	16.64
2		NaN	56717.97	64458.71	NaN	13.65
3		NaN	75526.00	71204.77	NaN	-5.72
4		NaN	67743.53	72477.54	NaN	6.99
5		NaN	73304.94	30456.61	NaN	-58.45
6		NaN	77405.65	NaN	NaN	NaN
7		NaN	83720.94	NaN	NaN	NaN
8		42397.56	75443.96	NaN	77.94	NaN
9		57001.42	66632.31	NaN	16.90	NaN
10		61404.85	71723.98	NaN	16.81	NaN
11		56636.08	68317.98	NaN	20.63	NaN
12		63949.58	73050.40	NaN	14.23	NaN

```

In [145]: non_zero_sales = sales[sales > 0]
min_sale = non_zero_sales.min()
max_sale = sales.max()

increase_pct = ((max_sale - min_sale) / min_sale) * 100
print(f"Sales increased by {increase_pct:.2f}% from the lowest (non-zero) to the highest day.")

```

Sales increased by 352.21% from the lowest (non-zero) to the highest day.

```

In [149]: import pandas as pd

# Load and prepare data
data = pd.read_csv("daily_sales_23_25.csv", header=None)
data.columns = [
    'period_no', 'week_no', 'date', 'store_no', 'store_name', 'total_revenue',
    'projected_psa', 'last_week_revenue', 'last_week_pct_change',
    'last_year_revenue', 'last_year_pct_change', 'lunch_sales', 'lunch_pct',
    'labor_pct', 'scheduled_pct', 'labor_cost', 'scheduled_hours',
    'actual_hours', 'hours_difference', 'dpmh', 'customer_count',

```

```

    'avg_ticket', 'status'
]

# Clean data
data['date'] = pd.to_datetime(data['date'], errors='coerce')
data['total_revenue'] = pd.to_numeric(data['total_revenue'], errors='coerce')
data = data.dropna(subset=['date', 'total_revenue'])

# Add year and month
data['year'] = data['date'].dt.year
data['month'] = data['date'].dt.month

### PART 1: 2023 vs 2024 comparison (common months only)
months_2023 = set(data[data['year'] == 2023]['month'])
months_2024 = set(data[data['year'] == 2024]['month'])
common_23_24 = sorted(months_2023 & months_2024)

data_23_24 = data[data['month'].isin(common_23_24) & data['year'].isin([2023, 2024])]
summary_23_24 = data_23_24.groupby('year')['total_revenue'].agg(['mean', 'sum'])
summary_23_24['Increase %'] = summary_23_24['mean'].pct_change().fillna(0) * 100
summary_23_24['Increase %'] = summary_23_24['Increase %'].round(2)
summary_23_24.columns = ['Year', 'Average Sale', 'Total Revenue', 'Increase %']

print("2023 vs 2024 Comparison (Same Months):")
print(summary_23_24)

### PART 2: 2024 vs 2025 comparison (up to latest month in 2025)
latest_month_2025 = data[data['year'] == 2025]['month'].max()
data_24_25 = data[(data['month'] <= latest_month_2025) & data['year'].isin([2024, 2025])]
summary_24_25 = data_24_25.groupby('year')['total_revenue'].agg(['mean', 'sum'])
summary_24_25['Increase %'] = summary_24_25['mean'].pct_change().fillna(0) * 100
summary_24_25['Increase %'] = summary_24_25['Increase %'].round(2)
summary_24_25.columns = ['Year', 'Average Sale', 'Total Revenue', 'Increase %']

print("\n2024 vs 2025 Comparison (Up to Month", latest_month_2025, "):")
print(summary_24_25)

```

2023 vs 2024 Comparison (Same Months):

	Year	Average Sale	Total Revenue	Increase %
0	2023	1927.33	281389.49	0.00
1	2024	2321.36	355168.63	20.44

2024 vs 2025 Comparison (Up to Month 5):

	Year	Average Sale	Total Revenue	Increase %
0	2024	2174.92	330587.62	0.00
1	2025	2279.48	305449.88	4.81

In []: