

# Data Analysis of Supplier Defects and Downtime

## Purpose

The purpose of this analysis is to gain insights into the defect quantities and downtime associated with different suppliers, materials, and plants. By understanding the patterns and correlations between these variables, the analysis aims to identify key areas of improvement to optimize quality control, reduce downtime, and improve supplier performance.

```
import pandas as pd
import numpy as np
import plotly.express as px
import matplotlib.pyplot as plt
import seaborn as sns

import pandas as pd

# Read all sheets from a single Excel file
file_path = "D:\Programs\Smart\my\Data analysis\DEPI\Final Project\
cleaned un-joined.xlsx" # Replace with your actual file path

# Assuming each sheet is named after the table name
dplant = pd.read_excel(file_path, sheet_name='DPlant')
dvendor = pd.read_excel(file_path, sheet_name='DVendor')
dcategory = pd.read_excel(file_path, sheet_name='DCategory')
dmaterial = pd.read_excel(file_path, sheet_name='DMaterial')
ddefect = pd.read_excel(file_path, sheet_name='DDefect')
ddefect_type = pd.read_excel(file_path, sheet_name='DDefect Type')
```

```
#calendar = pd.read_excel(file_path, sheet_name='Calendar')
supplier_quality = pd.read_excel(file_path, sheet_name='Supplier
Quality')
```

```
<>:4: SyntaxWarning:
```

```
invalid escape sequence '\P'
```

```
<>:4: SyntaxWarning:
```

```
invalid escape sequence '\P'
```

```
C:\Users\A\AppData\Local\Temp\ipykernel_10100\589000191.py:4:
SyntaxWarning:
```

```
invalid escape sequence '\P'
```

```
dplant.head()
```

	Plant ID	PlantDistrict	PlantState
0	1	GrandRapids	MI
1	2	Milwaukee	WI
2	3	Springfield	IL
3	4	Chicago	IL
4	5	Indianapolis	IN

```
dvendor.head()
```

	Vendor ID	Vendor
0	1	Reddoit
1	2	Plustax
2	3	bamity
3	4	Quotelane
4	5	Viatom

```
supplier_quality.head()
```

	Date	Category ID	Plant ID	Vendor ID	Material ID	Defect
0	2014-12-31	2	16	2	2137	
1	2014-12-31	2	20	59	1439	
2	2014-12-31	2	2	46	607	
3	2014-12-31	1	1	16	1824	
4	2014-12-31	4	5	4	54	

	Material Type ID	Defect ID	Defect Qty	Downtime min
0	6	281	1	60
1	8	295	9	10
2	8	299	47	30
3	3	90	20009	218
4	2	25	1	75

```
# Merge the Supplier Quality table with DPlant and DVendor
supplier_data = pd.merge(supplier_quality, dplant, on='Plant ID',
how='left')
supplier_data = pd.merge(supplier_data, dvendor, on='Vendor ID',
how='left')
supplier_data = pd.merge(supplier_data, dcategory, on='Category ID',
how='left')
supplier_data = pd.merge(supplier_data, dmaterial, on='Material Type
ID', how='left')
supplier_data = pd.merge(supplier_data, ddefect, on='Defect ID',
how='left')
supplier_data = pd.merge(supplier_data, ddefect_type, on='Defect Type
ID', how='left')
#supplier_data = pd.merge(supplier_data, calendar, on='Date',
how='left')
```

supplier\_data

	Date	Category ID	Plant ID	Vendor ID	Material ID \
0	2014-12-31	2	16	2	2137
1	2014-12-31	2	20	59	1439
2	2014-12-31	2	2	46	607
3	2014-12-31	1	1	16	1824
4	2014-12-31	4	5	4	54
...	...	...	...	...	...
5945	2013-01-01	1	1	102	446
5946	2013-01-01	1	1	101	441
5947	2013-01-01	3	4	17	151
5948	2013-01-01	3	4	17	151
5949	2013-01-01	2	3	1	7

	Defect Type ID	Material Type ID	Defect ID	Defect Qty
Downtime min \				
0	3	6	281	1
60				
1	3	8	295	9
10				
2	3	8	299	47
30				
3	3	3	90	20009
218				
4	4	2	25	1
75				

```

...
...
5945      4      17      185      80
0
5946      4      17      183      102
0
5947      1      13      95      0
0
5948      1      13      105      0
0
5949      3      1      8      0
0

```

```

      PlantDistrict PlantState      Vendor      Category \
0      Cincinnati      OH      Plustax      Logistics
1      Bangor      MI      Zuntexon      Logistics
2      Milwaukee      WI      Tamcan      Logistics
3      GrandRapids      MI      ontotam      Electrical
4      Indianapolis      IN      Quotelane      Mechanicals
...
5945      GrandRapids      MI      Zamholdings      Electrical
5946      GrandRapids      MI      Itdom      Electrical
5947      Chicago      IL      Trio-dax      Materials & Components
5948      Chicago      IL      Trio-dax      Materials & Components
5949      Springfield      IL      Reddoit      Logistics

```

```

      Material Type      Defect Defect Type      Sort
0      Controllers      Warped      Impact      2
1      Electrolytes      Deformed Parts      Impact      2
2      Electrolytes      Short Walls      Impact      2
3      Carton      Printing Defects      Impact      2
4      Film      Misc      Rejected      1
...
5945      Packaging      String Defects      Rejected      1
5946      Packaging      No Docs      Rejected      1
5947      Glass      Bad Bearings      No Impact      3
5948      Glass      Wrong Labeling      No Impact      3
5949      Corrugate      Overlapping Seam      Impact      2

```

```
[5950 rows x 18 columns]
```

```
# Drop all columns ending with 'ID' from the DataFrame
```

```
df_cleaned =
supplier_data.drop(supplier_data.filter(regex='ID$').columns, axis=1)
```

```
# Display the cleaned DataFrame
```

```
df_cleaned
df=df_cleaned
df
```

	Date	Defect	Qty	Downtime min	PlantDistrict	PlantState	\
0	2014-12-31		1	60	Cincinnati	OH	
1	2014-12-31		9	10	Bangor	MI	
2	2014-12-31		47	30	Milwaukee	WI	
3	2014-12-31		20009	218	GrandRapids	MI	
4	2014-12-31		1	75	Indianapolis	IN	
...	...		...	...	...	...	
5945	2013-01-01		80	0	GrandRapids	MI	
5946	2013-01-01		102	0	GrandRapids	MI	
5947	2013-01-01		0	0	Chicago	IL	
5948	2013-01-01		0	0	Chicago	IL	
5949	2013-01-01		0	0	Springfield	IL	
	Vendor			Category	Material	Type	
Defect \							
0	Plustax			Logistics	Controllers		
Warped							
1	Zuntexon			Logistics	Electrolytes	Deformed	
Parts							
2	Tamcan			Logistics	Electrolytes	Short	
Walls							
3	ontotam			Electrical	Carton	Printing	
Defects							
4	Quotelane			Mechanicals	Film		
Misc							
...	...			...	...		
...							
5945	Zamholdings			Electrical	Packaging	String	
Defects							
5946	Itdom			Electrical	Packaging	No	
Docs							
5947	Trio-dax	Materials & Components			Glass	Bad	
Bearings							
5948	Trio-dax	Materials & Components			Glass	Wrong	
Labeling							
5949	Reddoit			Logistics	Corrugate	Overlapping	
Seam							
	Defect	Type	Sort				
0		Impact	2				
1		Impact	2				
2		Impact	2				
3		Impact	2				
4		Rejected	1				
...		...	...				
5945		Rejected	1				
5946		Rejected	1				
5947	No	Impact	3				
5948	No	Impact	3				
5949		Impact	2				

```
[5950 rows x 11 columns]
```

```
# Check for missing values
```

```
df.isnull().sum()
```

```
Date          0
Defect Qty     0
Downtime min   0
PlantDistrict  0
PlantState     0
Vendor         0
Category       0
Material Type  0
Defect         0
Defect Type    0
Sort          0
dtype: int64
```

```
df.describe()
```

	Date	Defect Qty	Downtime min
Sort			
count	5950	5950.000000	5950.000000
5950.000000			
mean	2014-01-24 09:31:53.142857216	9166.254790	23.235462
2.124370			
min	2013-01-01 00:00:00	0.000000	0.000000
1.000000			
25%	2013-07-21 06:00:00	9.000000	0.000000
1.000000			
50%	2014-02-08 00:00:00	445.000000	0.000000
2.000000			
75%	2014-08-26 00:00:00	5195.750000	20.000000
3.000000			
max	2014-12-31 00:00:00	487008.000000	999.000000
3.000000			
std	NaN	30590.589573	76.354891
0.814432			

```
# Convert columns to appropriate data types (if necessary)
```

```
supplier_data['Date'] = pd.to_datetime(supplier_data['Date'])
```

# Vendor Analysis

## Top 10 vendor with high defects

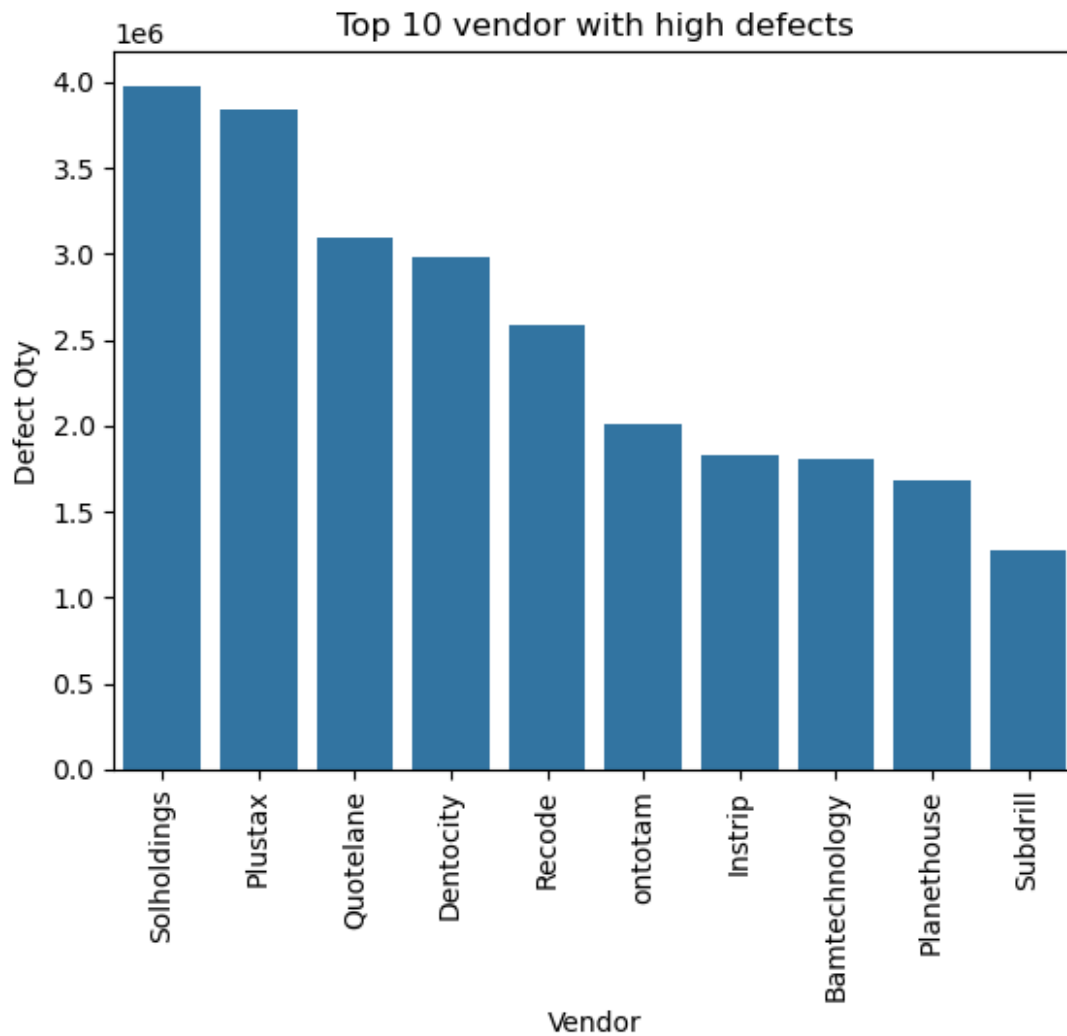
```
# Group by Vendor and calculate total defects and downtime
vendor_analysis = supplier_data.groupby('Vendor')[['Defect Qty',
'Downtime min']].sum().reset_index()
# Sort vendor_analysis in ascending order by 'Defect Qty' and
'Downtime min'
vendor_analysis_sorted = vendor_analysis.sort_values(by=['Defect Qty',
'Downtime min'], ascending=False)

# Display the sorted DataFrame
vendor_analysis_sorted.head(10)
```

	Vendor	Defect Qty	Downtime min
188	Solholdings	3977962	2275
153	Plustax	3836303	10270
157	Quotelane	3095198	5831
34	Dentocity	2982348	3088
162	Recode	2589319	4215
291	ontotam	2005374	3327
98	Instrip	1828614	3088
5	Bamtechnology	1810404	1030
146	Planethouse	1681683	3382
204	Subdrill	1273832	2221

```
# Plot total defects by vendor
import matplotlib.pyplot as plt
import seaborn as sns

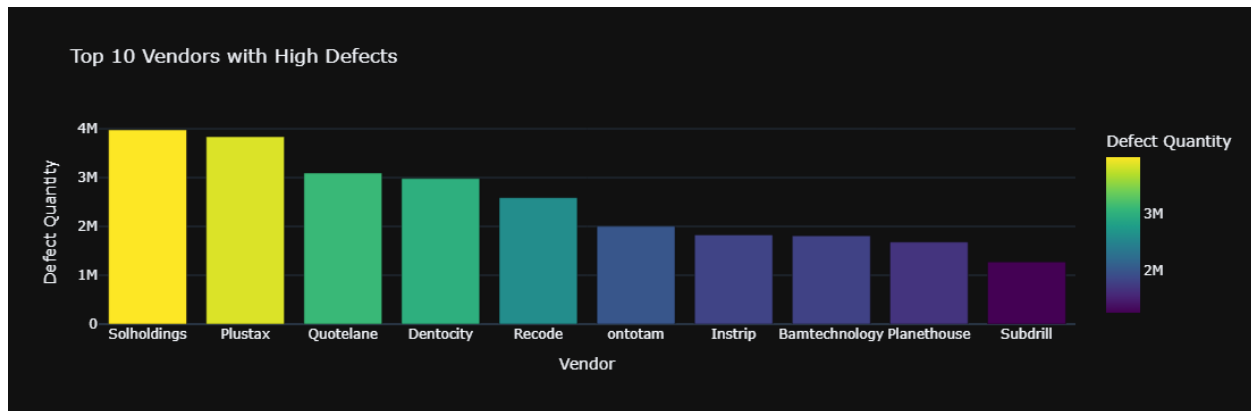
sns.barplot(data=vendor_analysis_sorted.head(10), x='Vendor',
y='Defect Qty')
plt.xticks(rotation=90)
plt.title('Top 10 vendor with high defects')
plt.show()
```



```
# Create a bar chart for top 10 vendors with high defect quantities
using plotly
fig = px.bar(vendor_analysis_sorted.head(10),
             x='Vendor',
             y='Defect Qty',
             title='Top 10 Vendors with High Defects',
             labels={'Defect Qty': 'Defect Quantity'},
             template='plotly_dark',
             color='Defect Qty',
             color_continuous_scale='Viridis')

# Display the plot
fig.show()
```





## vendors with a Zero Defect Qty

```
# Filter vendors with 'Defect Qty' equal to 0
vendors_with_zero_defects = vendor_analysis[vendor_analysis['Defect Qty'] == 0]
```

```
# Display the filtered DataFrame
vendors_with_zero_defects.count()
```

```
Vendor      54
Defect Qty   54
Downtime min 54
dtype: int64
```

```
# Get a list of vendors with 'Defect Qty' equal to 0
vendors_with_zero_defects_list =
vendors_with_zero_defects['Vendor'].tolist()
```

```
# Display the list
vendors_with_zero_defects_list
```

```
['Bamgeohigh',
 'Canace',
 'Canphase',
 'Condexon',
 'Daltcare',
 'Daltron',
 'Danfan',
 'Dentoelectrics',
 'Dingfax',
 'Dongbase',
 'Finlux',
 'Fixcan',
 'Geoit',
 'Goldenlex',
 'Goodtechno',
 'Hayholding',
 'Indigoice',
```

```
'Iselectrics',  
'Ittam',  
'Joyzim',  
'Kantone',  
'Konknix',  
'Latgotrax',  
'Linetone',  
'Matsanice',  
'Planetware',  
'Pluslam',  
'Scothouse',  
'Silver-line',  
'Solohex',  
'Sonlex',  
'Statway',  
'Streetplus',  
'Templax',  
'Toncode',  
'Unatrax',  
'Viacane',  
'Vivazim',  
'Warecity',  
'Y-strip',  
'Zamlane',  
'Zathlane',  
'Zerlane',  
'Zimex',  
'Zumcane',  
'Zuntone',  
'betaity',  
'dripkix',  
'saltlab',  
'san-plex',  
'stripzim',  
'tinron',  
'wareholding',  
'zen-holdings']
```

it's seem that thre are 54 vendor with no defect's

```
df.columns
```

```
Index(['Date', 'Defect Qty', 'Downtime min', 'PlantDistrict',  
      'PlantState',  
      'Vendor', 'Category', 'Material Type', 'Defect', 'Defect Type',  
      'Sort'],  
      dtype='object')
```

# Category Analysis

```
# Group by Category and calculate total defects and downtime
category_analysis = supplier_data.groupby('Category')[['Defect Qty',
'Downtime min']].sum().reset_index()
```

```
# Display the analysis results
```

```
category_analysis_sorted = category_analysis.sort_values(by='Defect
Qty', ascending=False)
```

```
category_analysis_sorted
```

	Category	Defect Qty	Downtime min
4	Mechanicals	16989072	34208
5	Packaging	15705616	29326
2	Logistics	13122995	55849
3	Materials & Components	3714589	4946
1	Goods & Services	3356864	11964
0	Electrical	1650080	1958

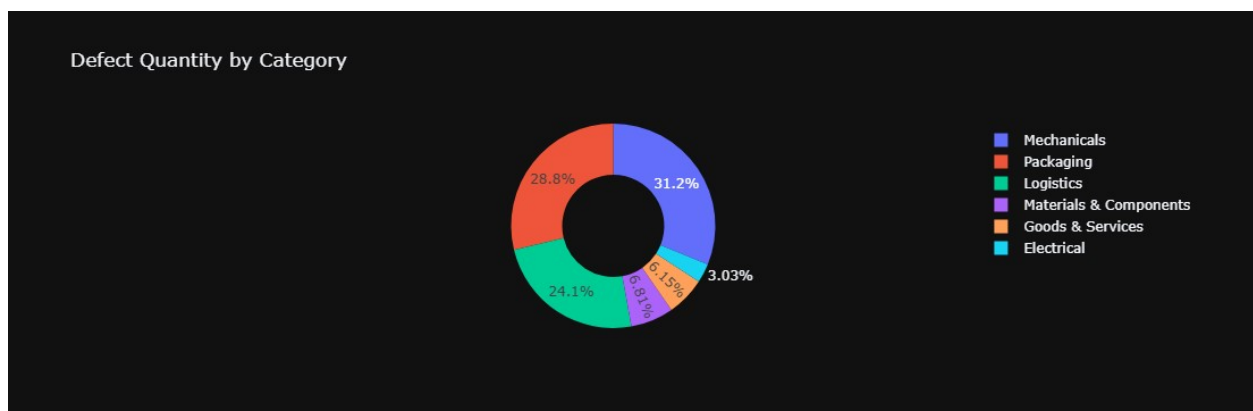
```
import plotly.express as px
```

```
# Create a pie chart using plotly to visualize the defect quantity by
category
```

```
fig = px.pie(category_analysis_sorted,
              values='Defect Qty',
              names='Category',
              title='Defect Quantity by Category',
              hole=0.5,
              template='plotly_dark')
```

```
# Show the pie chart
```

```
fig.show()
```



```
df
```

	Date	Defect Qty	Downtime min	PlantDistrict	PlantState	\
0	2014-12-31	1	60	Cincinnati	OH	

1	2014-12-31	9	10	Bangor	MI
2	2014-12-31	47	30	Milwaukee	Wi
3	2014-12-31	20009	218	GrandRapids	MI
4	2014-12-31	1	75	Indianapolis	IN
...	...	...	...	...	...
5945	2013-01-01	80	0	GrandRapids	MI
5946	2013-01-01	102	0	GrandRapids	MI
5947	2013-01-01	0	0	Chicago	IL
5948	2013-01-01	0	0	Chicago	IL
5949	2013-01-01	0	0	Springfield	IL

	Vendor	Category	Material	Type
Defect \				
0	Plustax	Logistics	Controllers	
Warped				
1	Zuntexon	Logistics	Electrolytes	Deformed
Parts				
2	Tamcan	Logistics	Electrolytes	Short
Walls				
3	ontotam	Electrical	Carton	Printing
Defects				
4	Quotelane	Mechanicals	Film	
Misc				
...	...	...	...	...
...				
5945	Zamholdings	Electrical	Packaging	String
Defects				
5946	Itdom	Electrical	Packaging	No
Docs				
5947	Trio-dax	Materials & Components	Glass	Bad
Bearings				
5948	Trio-dax	Materials & Components	Glass	Wrong
Labeling				
5949	Reddoit	Logistics	Corrugate	Overlapping
Seam				

	Defect	Type	Sort
0	Impact		2
1	Impact		2
2	Impact		2
3	Impact		2
4	Rejected		1
...	...		...
5945	Rejected		1
5946	Rejected		1
5947	No Impact		3
5948	No Impact		3
5949	Impact		2

[5950 rows x 11 columns]

# plant Analysis

```
# Group by Plant and calculate total defects and downtime
plant_analysis = supplier_data.groupby('PlantDistrict')[['Defect Qty',
'Downtime min']].sum().reset_index()

# Sort the results by total defects
plant_analysis_sorted = plant_analysis.sort_values(by='Defect Qty',
ascending=False)
```

```
# Display the analysis results
plant_analysis_sorted
```

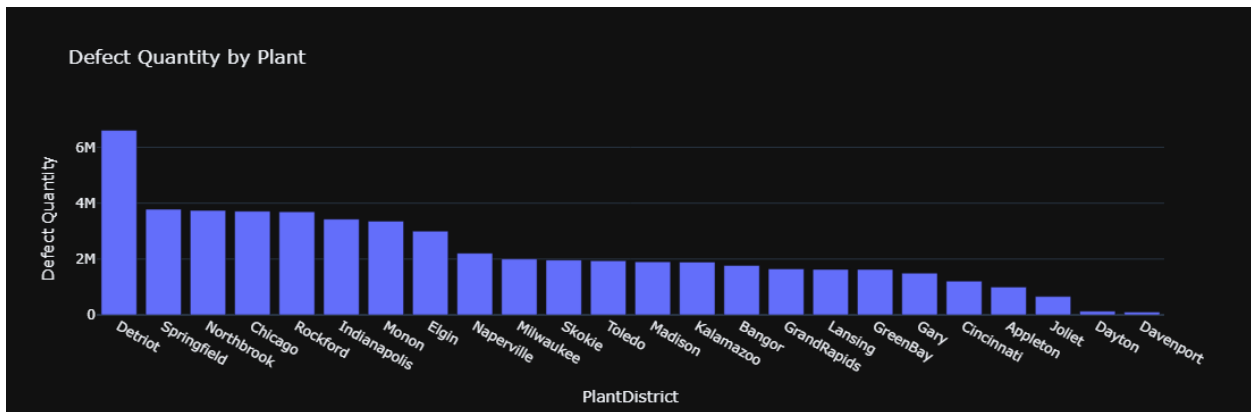
	PlantDistrict	Defect Qty	Downtime min
6	Detriot	6610077	11428
22	Springfield	3784005	17296
19	Northbrook	3740202	8175
2	Chicago	3714589	4946
20	Rockford	3697133	5799
11	Indianapolis	3431509	11305
17	Monon	3356864	11964
7	Elgin	3000039	1795
18	Naperville	2211497	4601
16	Milwaukee	2003160	1127
21	Skokie	1968165	9052
23	Toledo	1938406	3893
15	Madison	1904438	6633
13	Kalamazoo	1893353	3198
1	Bangor	1773241	9531
9	GrandRapids	1650080	1958
14	Lansing	1631197	465
10	GreenBay	1629397	3580
8	Gary	1497006	920
3	Cincinnati	1209796	14745
0	Appleton	997961	0
12	Joliet	660243	3924
5	Dayton	131605	191
4	Davenport	105253	1725

```
import plotly.express as px
```

```
# Create a bar chart for defect quantity by plant
```

```
fig = px.bar(plant_analysis_sorted,
             x='PlantDistrict',
             y='Defect Qty',
             title='Defect Quantity by Plant',
             labels={'Defect Qty': 'Defect Quantity'},
             template='plotly_dark')
```

```
# Show the plot
fig.show()
```



```
# Group by Material Type and calculate total defects and downtime
material_type_analysis = supplier_data.groupby('Material Type')
[['Defect Qty', 'Downtime min']].sum().reset_index()

# Sort the results by total defect quantity
material_type_analysis_sorted =
material_type_analysis.sort_values(by='Defect Qty', ascending=False)

# Display the analysis results
material_type_analysis_sorted
```

	Material Type	Defect Qty	Downtime min
18	Raw Materials	13218621	23568
11	Labels	8258962	7017
1	Carton	7923317	12869
8	Film	7017127	8608
3	Controllers	4171910	8555
7	Electrolytes	3023458	5975
13	Molds	2296332	4263
4	Corrugate	2208438	52726
10	Hardware	1555175	3268
2	Composites	1258623	758
0	Batteries	1160787	8234
17	Pump	907088	340
6	Drives	539076	390
14	Motors	538985	397
9	Glass	254946	583
12	Mechanicals	83134	495
19	Tape	64153	205
5	Crates	33381	0
15	Packaging	19891	0
20	Valves	5696	0

```

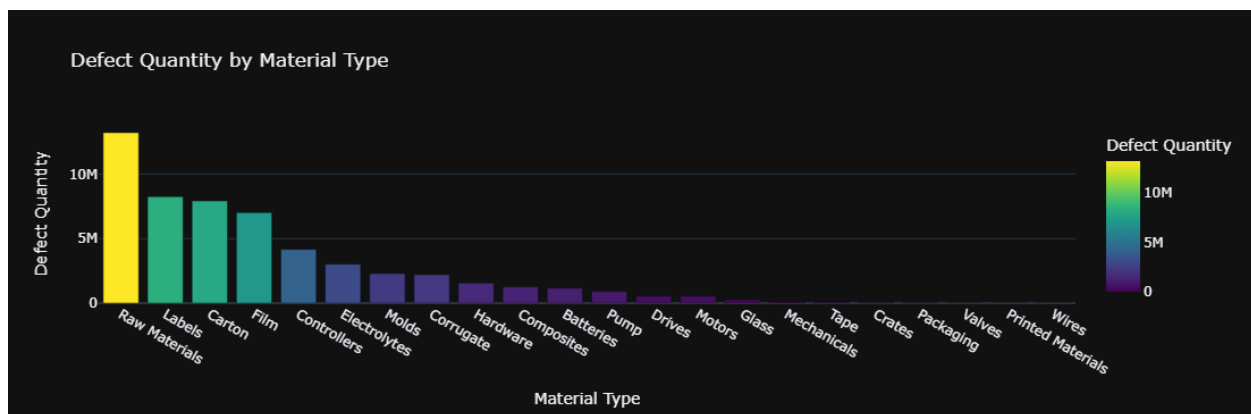
16 Printed Materials      116      0
21 Wires                  0      0

import plotly.express as px

# Create a bar chart for defect quantity by material type
fig = px.bar(material_type_analysis_sorted,
             x='Material Type',
             y='Defect Qty',
             title='Defect Quantity by Material Type',
             labels={'Defect Qty': 'Defect Quantity'},
             template='plotly_dark',
             color='Defect Qty',
             color_continuous_scale='Viridis') # Choose a color
palette

# Display the plot
fig.show()

```



```

df

```

	Date	Defect Qty	Downtime min	PlantDistrict	PlantState	\
0	2014-12-31	1	60	Cincinnati	OH	
1	2014-12-31	9	10	Bangor	MI	
2	2014-12-31	47	30	Milwaukee	Wi	
3	2014-12-31	20009	218	GrandRapids	MI	
4	2014-12-31	1	75	Indianapolis	IN	
...	...	...	...	...	...	...
5945	2013-01-01	80	0	GrandRapids	MI	
5946	2013-01-01	102	0	GrandRapids	MI	
5947	2013-01-01	0	0	Chicago	IL	
5948	2013-01-01	0	0	Chicago	IL	
5949	2013-01-01	0	0	Springfield	IL	

Defect	Vendor	Category	Material Type
0	Plustax	Logistics	Controllers

```

Warped
1      Zuntexon      Logistics  Electrolytes  Deformed
Parts
2      Tamcan      Logistics  Electrolytes  Short
Walls
3      ontotam      Electrical  Carton  Printing
Defects
4      Quotelane      Mechanicals  Film
Misc
...      ...      ...      ...
...
5945  Zamholdings      Electrical  Packaging  String
Defects
5946      Itdom      Electrical  Packaging  No
Docs
5947      Trio-dax  Materials & Components  Glass  Bad
Bearings
5948      Trio-dax  Materials & Components  Glass  Wrong
Labeling
5949      Reddoit      Logistics  Corrugate  Overlapping
Seam

      Defect Type  Sort
0      Impact      2
1      Impact      2
2      Impact      2
3      Impact      2
4      Rejected      1
...      ...      ...
5945      Rejected      1
5946      Rejected      1
5947      No Impact      3
5948      No Impact      3
5949      Impact      2

[5950 rows x 11 columns]

```

## Defect Type Analysis

```

# Group by Defect Type and calculate total defect quantity and
downtime
defect_analysis = supplier_data.groupby('Defect Type')[['Defect Qty',
'Downtime min']].sum().reset_index()

# Sort the results by total defect quantity
defect_analysis_sorted = defect_analysis.sort_values(by='Defect Qty',
ascending=False)

```



```
# Display the analysis results
```

```
defect_analysis_sorted
```

	Defect Type	Defect Qty	Downtime min
2	Rejected	19369349	28435
0	Impact	19301562	104677
1	No Impact	15868305	5139

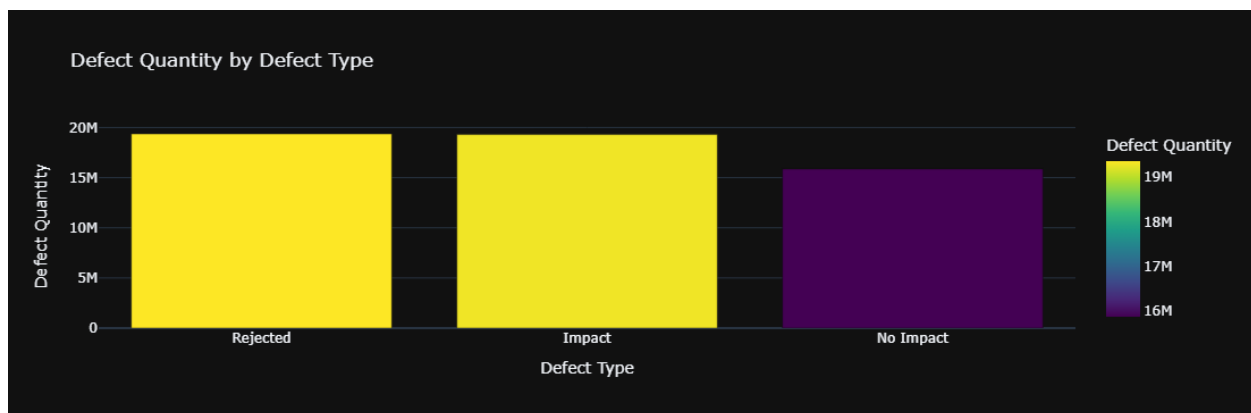
```
import plotly.express as px
```

```
# Create a bar chart for defect quantity by defect type
```

```
fig = px.bar(defect_analysis_sorted,  
             x='Defect Type',  
             y='Defect Qty',  
             title='Defect Quantity by Defect Type',  
             labels={'Defect Qty': 'Defect Quantity'},  
             template='plotly_dark',  
             color='Defect Qty',  
             color_continuous_scale='Viridis')
```

```
# Display the plot
```

```
fig.show()
```



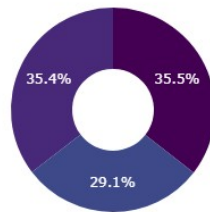
```
# Create a pie chart for defect quantity by defect type
```

```
fig = px.pie(defect_analysis_sorted,  
            values='Defect Qty',  
            names='Defect Type',  
            title='Defect Quantity Distribution by Defect Type',  
            hole=0.4, # Creates a donut chart  
            color_discrete_sequence=px.colors.sequential.Viridis)
```

```
# Display the plot
```

```
fig.show()
```

Defect Quantity Distribution by Defect Type



■ Rejected  
 ■ Impact  
 ■ No Impact

df

	Date	Defect Qty	Downtime min	PlantDistrict	PlantState	\
0	2014-12-31	1	60	Cincinnati	OH	
1	2014-12-31	9	10	Bangor	MI	
2	2014-12-31	47	30	Milwaukee	Wi	
3	2014-12-31	20009	218	GrandRapids	MI	
4	2014-12-31	1	75	Indianapolis	IN	
...	...	...	...	...	...	...
5945	2013-01-01	80	0	GrandRapids	MI	
5946	2013-01-01	102	0	GrandRapids	MI	
5947	2013-01-01	0	0	Chicago	IL	
5948	2013-01-01	0	0	Chicago	IL	
5949	2013-01-01	0	0	Springfield	IL	

	Vendor	Category	Material	Type
Defect \				
0	Plustax	Logistics	Controllers	
Warped				
1	Zuntexon	Logistics	Electrolytes	Deformed
Parts				
2	Tamcan	Logistics	Electrolytes	Short
Walls				
3	ontotam	Electrical	Carton	Printing
Defects				
4	Quotelane	Mechanicals	Film	
Misc				
...	...	...	...	...
...				
5945	Zamholdings	Electrical	Packaging	String
Defects				
5946	Itdom	Electrical	Packaging	No
Docs				
5947	Trio-dax	Materials & Components	Glass	Bad
Bearings				
5948	Trio-dax	Materials & Components	Glass	Wrong
Labeling				

5949	Reddoit	Logistics	Corrugate	Overlapping
Seam				

	Defect Type	Sort
0	Impact	2
1	Impact	2
2	Impact	2
3	Impact	2
4	Rejected	1
...	...	...
5945	Rejected	1
5946	Rejected	1
5947	No Impact	3
5948	No Impact	3
5949	Impact	2

[5950 rows x 11 columns]

## Top 10 Defect Quantities by Defects

```
# Group the data by Defect to get total defect quantities and
associated downtime
defect_analysis = supplier_data.groupby('Defect')[['Defect Qty',
'Downtime min']].sum().reset_index()

# Sort by the total defect quantity for better visualization
defect_analysis_sorted = defect_analysis.sort_values(by='Defect Qty',
ascending=False).reset_index(drop=True)

# Display the top 10 results without an extra index
defect_analysis_sorted=defect_analysis_sorted.head(10)
defect_analysis_sorted
```

	Defect	Defect Qty	Downtime min
0	Misc	4583814	5881
1	Not Certified	3686660	60
2	Warped	2866749	7050
3	Wrong Shade of Color	1997933	824
4	Printing Defects	1646378	1760
5	Bad Parameters	1454232	180
6	Out of Spec	1428879	3894
7	Foreign Material	1115095	5005
8	Other	1083164	960
9	Loose Core	1035519	455

```
# Group by Vendor and Defect Type to see how defects are spread across
vendors
vendor_defect_analysis = supplier_data.groupby(['Vendor', 'Defect
```

```
Type']][['Defect Qty']].sum().reset_index()
```

```
# Display the top entries for insight
```

```
vendor_defect_analysis.sort_values(by='Defect Qty', ascending=False)
```

	Vendor	Defect Type	Defect Qty
244	Plustax	Impact	3267202
314	Solholdings	Rejected	2020209
234	Planethouse	Rejected	1540204
312	Solholdings	Impact	1423984
263	Recode	Impact	1244949
..	...	...	...
400	Viacane	No Impact	0
288	Saohow	Rejected	0
86	Duoflex	Impact	0
317	Solozap	Impact	0
442	Zerlane	No Impact	0

```
[548 rows x 3 columns]
```

```
df
```

	Date	Defect Qty	Downtime min	PlantDistrict	PlantState	\
0	2014-12-31	1	60	Cincinnati	OH	
1	2014-12-31	9	10	Bangor	MI	
2	2014-12-31	47	30	Milwaukee	WI	
3	2014-12-31	20009	218	GrandRapids	MI	
4	2014-12-31	1	75	Indianapolis	IN	
...	...	...	...	...	...	
5945	2013-01-01	80	0	GrandRapids	MI	
5946	2013-01-01	102	0	GrandRapids	MI	
5947	2013-01-01	0	0	Chicago	IL	
5948	2013-01-01	0	0	Chicago	IL	
5949	2013-01-01	0	0	Springfield	IL	

	Vendor	Category	Material	Type
Defect \				
0	Plustax	Logistics	Controllers	
Warped				
1	Zuntexon	Logistics	Electrolytes	Deformed
Parts				
2	Tamcan	Logistics	Electrolytes	Short
Walls				
3	ontotam	Electrical	Carton	Printing
Defects				
4	Quotelane	Mechanicals	Film	
Misc				
...	...	...	...	...
...				

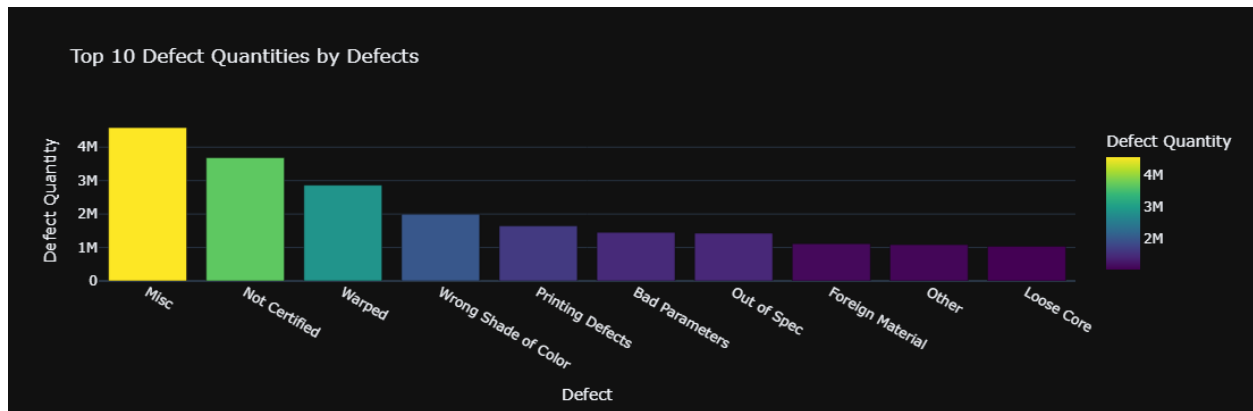
5945	Zamholdings	Electrical	Packaging	String
5946	Itdom	Electrical	Packaging	No
5947	Trio-dax	Materials & Components	Glass	Bad
5948	Trio-dax	Materials & Components	Glass	Wrong
5949	Reddoit	Logistics	Corrugate	Overlapping

	Defect Type	Sort
0	Impact	2
1	Impact	2
2	Impact	2
3	Impact	2
4	Rejected	1
...	...	...
5945	Rejected	1
5946	Rejected	1
5947	No Impact	3
5948	No Impact	3
5949	Impact	2

[5950 rows x 11 columns]

```
# Create a bar chart to show total defect quantities by defect type
fig = px.bar(defect_analysis_sorted,
             x='Defect',
             y='Defect Qty',
             title='Top 10 Defect Quantities by Defects',
             labels={'Defect Qty': 'Defect Quantity'},
             template='plotly_dark',
             color='Defect Qty',
             color_continuous_scale='Viridis')

fig.update_layout(xaxis={'categoryorder': 'total descending'})
# Display the plot
fig.show()
```



df

	Date	Defect Qty	Downtime min	PlantDistrict	PlantState	\
0	2014-12-31	1	60	Cincinnati	OH	
1	2014-12-31	9	10	Bangor	MI	
2	2014-12-31	47	30	Milwaukee	WI	
3	2014-12-31	20009	218	GrandRapids	MI	
4	2014-12-31	1	75	Indianapolis	IN	
...	...	...	...	...	...	...
5945	2013-01-01	80	0	GrandRapids	MI	
5946	2013-01-01	102	0	GrandRapids	MI	
5947	2013-01-01	0	0	Chicago	IL	
5948	2013-01-01	0	0	Chicago	IL	
5949	2013-01-01	0	0	Springfield	IL	

	Vendor	Category	Material	Type
Defect \				
0	Plustax	Logistics	Controllers	
Warped				
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Defects				
4	Quotelane	Mechanicals	Film	
Misc				
...	...	...	...	...
...				
5945	Zamholdings	Electrical	Packaging	String
Defects				
5946	Itdom	Electrical	Packaging	No
Docs				
5947	Trio-dax	Materials & Components	Glass	Bad
Bearings				
5948	Trio-dax	Materials & Components	Glass	Wrong
Labeling				

5949	Reddoit	Logistics	Corrugate	Overlapping
Seam				

	Defect Type	Sort
0	Impact	2
1	Impact	2
2	Impact	2
3	Impact	2
4	Rejected	1
...	...	...
5945	Rejected	1
5946	Rejected	1
5947	No Impact	3
5948	No Impact	3
5949	Impact	2

[5950 rows x 11 columns]

```
# Convert 'Date' column to datetime if not already
df['Date'] = pd.to_datetime(supplier_data['Date'])
```

```
# Group by month and year to analyze defects over time
df['Year'] = df['Date'].dt.year
df['Month'] = df['Date'].dt.month
```

```
# Group by year and month to calculate total defects and downtime
date_analysis = df.groupby(['Year', 'Month'])[['Defect Qty', 'Downtime min']].sum().reset_index()
```

```
# Display the analysis results
date_analysis
```

	Year	Month	Defect Qty	Downtime min
0	2013	1	1881043	5321
1	2013	2	1434261	5855
2	2013	3	1894519	7415
3	2013	4	3090509	5108
4	2013	5	2044767	2138
5	2013	6	2928678	4170
6	2013	7	1635252	3616
7	2013	8	712815	5762
8	2013	9	1881043	5321
9	2013	10	1528768	7249
10	2013	11	1800012	6021
11	2013	12	2085678	4295
12	2014	1	2911213	8259
13	2014	2	776423	5144
14	2014	3	1834428	6581
15	2014	4	2278627	4473
16	2014	5	1110777	3586

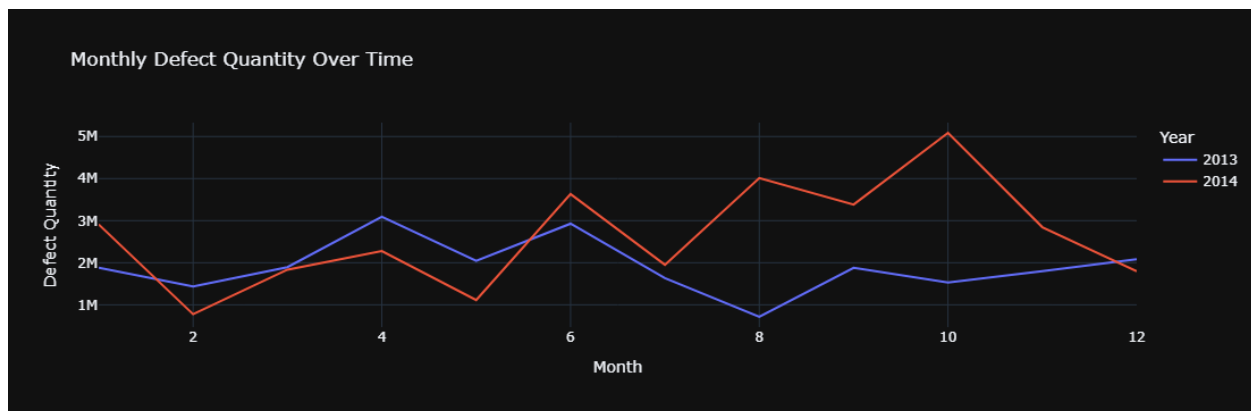
17	2014	6	3632208	7229
18	2014	7	1951816	3547
19	2014	8	4011608	5104
20	2014	9	3383591	5749
21	2014	10	5087899	12322
22	2014	11	2844996	7074
23	2014	12	1798285	6912

## Defect Quantities Over Time

```
import plotly.express as px

# Create a line chart for defect quantities over time
fig = px.line(date_analysis,
              x='Month',
              y='Defect Qty',
              color='Year',
              title='Monthly Defect Quantity Over Time',
              labels={'Defect Qty': 'Defect Quantity'},
              template='plotly_dark')

fig.show()
```

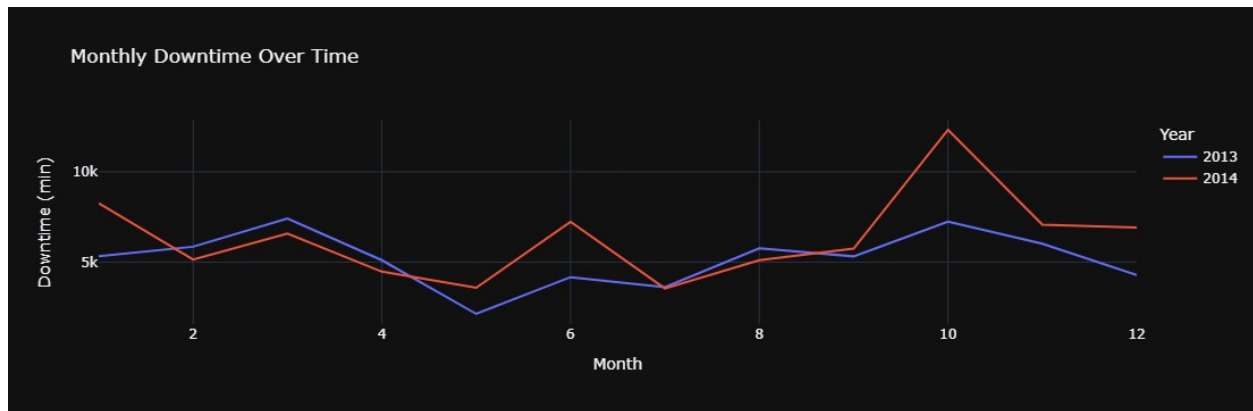


## Downtime Over Time

```
# Create a line chart for downtime over time
fig = px.line(date_analysis,
              x='Month',
              y='Downtime min',
              color='Year',
              title='Monthly Downtime Over Time',
              labels={'Downtime min': 'Downtime (min)'},
              template='plotly_dark')
```



```
fig.show()
```

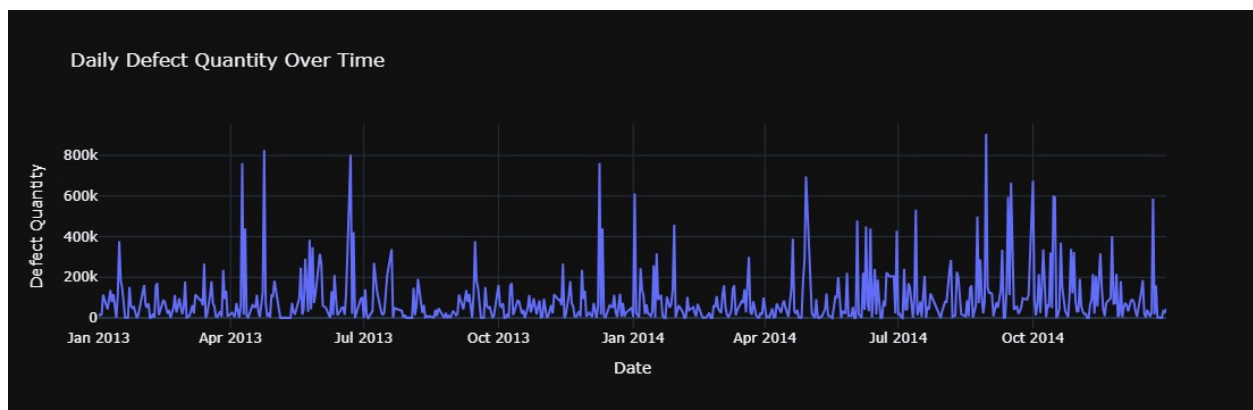


## Daily or Weekly Analysis

```
# Group by day to see daily trends
daily_analysis = supplier_data.groupby('Date')[['Defect Qty',
'Downtime min']].sum().reset_index()

# Create a line chart for daily defect quantities
fig = px.line(daily_analysis,
              x='Date',
              y='Defect Qty',
              title='Daily Defect Quantity Over Time',
              labels={'Defect Qty': 'Defect Quantity'},
              template='plotly_dark')

fig.show()
```



# Seasonal or Quarter Analysis

```
# Ensure the 'Date' column is in datetime format
supplier_data['Date'] = pd.to_datetime(supplier_data['Date'])

# Extract Year and Quarter
supplier_data['Year'] = supplier_data['Date'].dt.year
supplier_data['Quarter'] = supplier_data['Date'].dt.quarter

# Group by 'Year' and 'Quarter' for analysis
quarterly_analysis = supplier_data.groupby(['Year', 'Quarter'])
[ ['Defect Qty', 'Downtime min']].sum().reset_index()

# Display the analysis
quarterly_analysis.head()
```

	Year	Quarter	Defect Qty	Downtime min
0	2013	1	5209823	18591
1	2013	2	8063954	11416
2	2013	3	4229110	14699
3	2013	4	5414458	17565
4	2014	1	5522064	19984

```
import plotly.express as px

# Ensure 'Date' column is in datetime format
supplier_data['Date'] = pd.to_datetime(supplier_data['Date'])

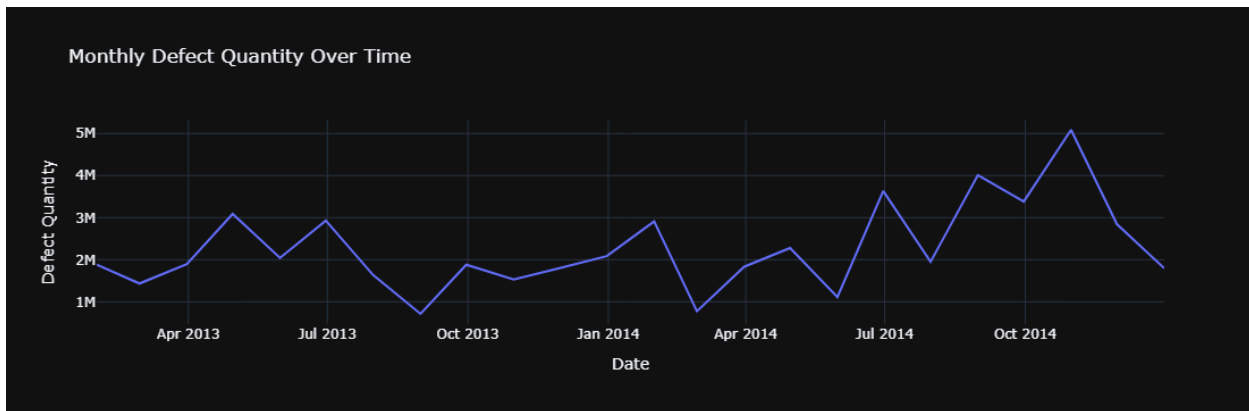
# Aggregate defect quantity by month
monthly_defects = supplier_data.resample('M',
on='Date').sum().reset_index()

# Create a line chart for defect quantity over time
fig = px.line(monthly_defects,
               x='Date',
               y='Defect Qty',
               title='Monthly Defect Quantity Over Time',
               labels={'Defect Qty': 'Defect Quantity'},
               template='plotly_dark')

# Display the line chart
fig.show()
```

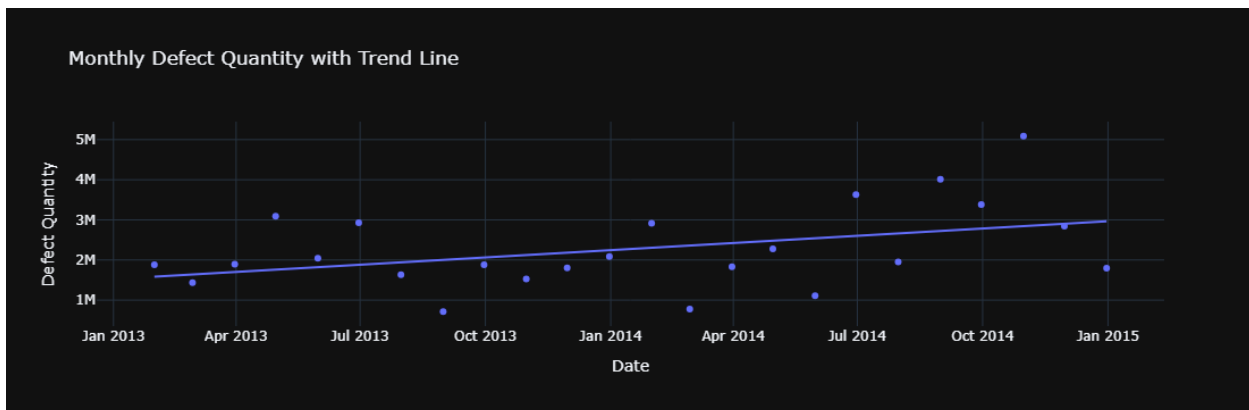
C:\Users\A\AppData\Local\Temp\ipykernel\_10100\884135220.py:7:  
FutureWarning:

'M' is deprecated and will be removed in a future version, please use  
'ME' instead.



```
# Create a scatter plot for defect quantity with trend line
fig = px.scatter(monthly_defects,
                 x='Date',
                 y='Defect Qty',
                 title='Monthly Defect Quantity with Trend Line',
                 labels={'Defect Qty': 'Defect Quantity'},
                 trendline='ols', # Add an Ordinary Least Squares
                                # (OLS) trendline
                 template='plotly_dark')

# Display the scatter plot
fig.show()
```



```
# Create 'Year' and 'Month' columns
supplier_data['Year'] = supplier_data['Date'].dt.year
supplier_data['Month'] = supplier_data['Date'].dt.month

# Group by month and year
monthly_defect_summary = supplier_data.groupby(['Year', 'Month'])
                                ['Defect Qty'].sum().reset_index()

# Create a heatmap to visualize defects over time
fig = px.density_heatmap(monthly_defect_summary,
                        x='Month',
```

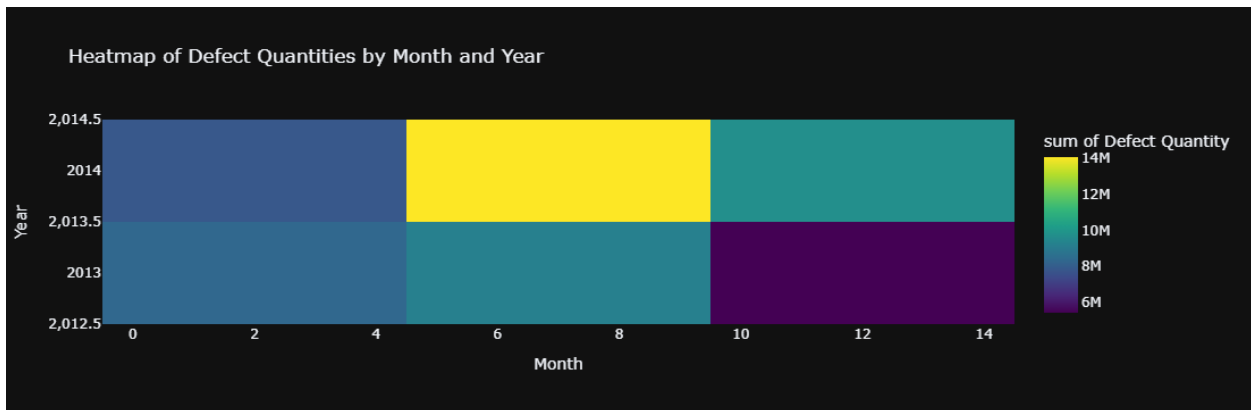
```

y='Year',
z='Defect Qty',
title='Heatmap of Defect Quantities by Month
and Year',

labels={'Defect Qty': 'Defect Quantity'},
template='plotly_dark',
color_continuous_scale='Viridis')

# Display the heatmap
fig.show()

```



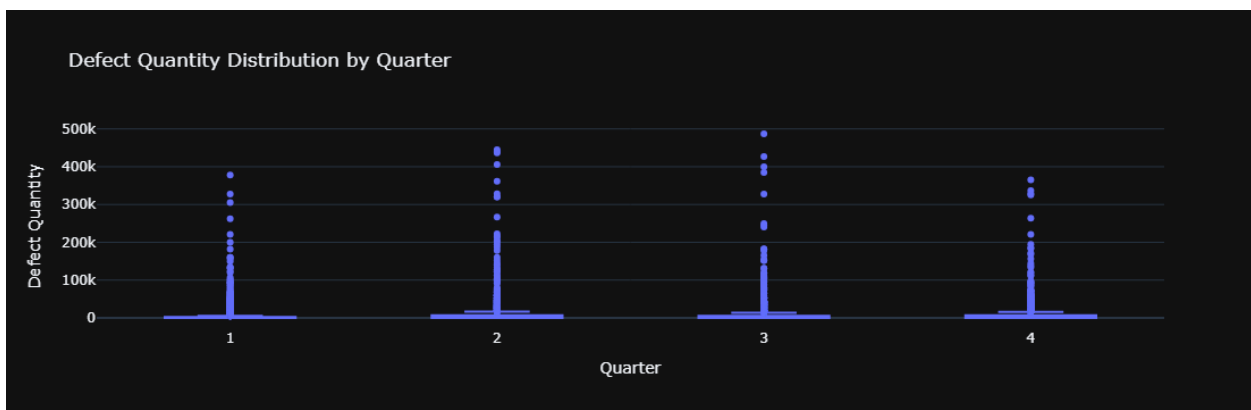
```

# Create a 'Quarter' column
supplier_data['Quarter'] = supplier_data['Date'].dt.quarter

# Create a box plot to visualize defect distribution by quarter
fig = px.box(supplier_data,
             x='Quarter',
             y='Defect Qty',
             title='Defect Quantity Distribution by Quarter',
             labels={'Defect Qty': 'Defect Quantity'},
             template='plotly_dark')

# Display the box plot
fig.show()

```



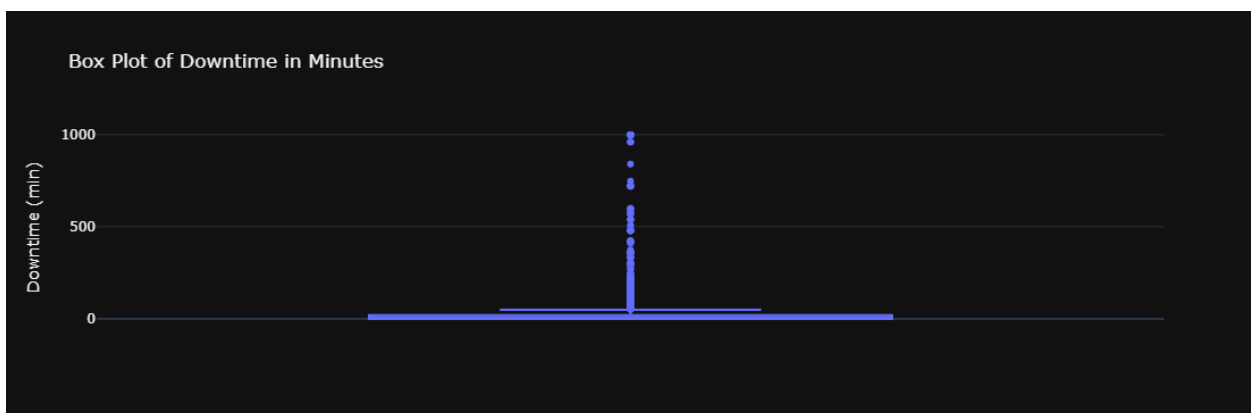
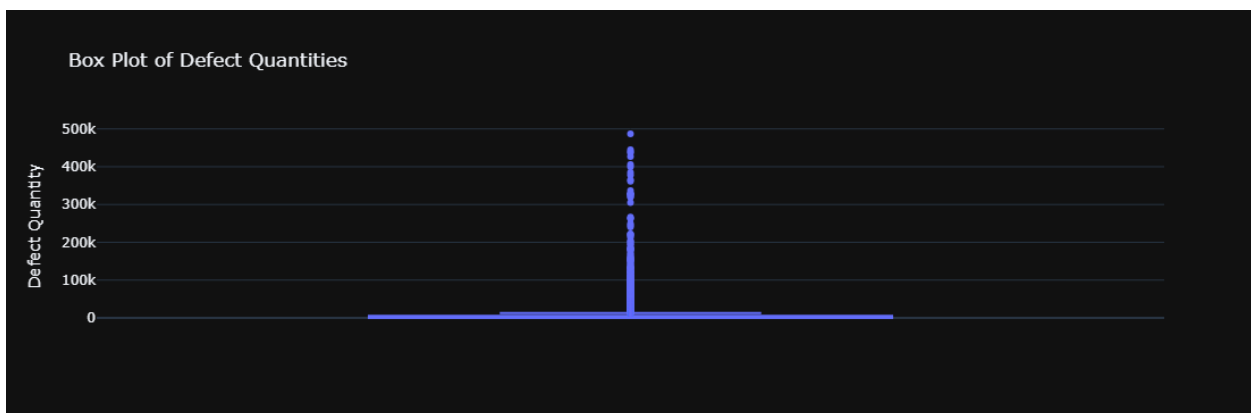
```
import plotly.express as px

# Create a box plot for 'Defect Qty'
fig_qty = px.box(supplier_data,
                  y='Defect Qty',
                  title='Box Plot of Defect Quantities',
                  labels={'Defect Qty': 'Defect Quantity'},
                  template='plotly_dark')

# Display the box plot for 'Defect Qty'
fig_qty.show()

# Create a box plot for 'Downtime min'
fig_downtime = px.box(supplier_data,
                       y='Downtime min',
                       title='Box Plot of Downtime in Minutes',
                       labels={'Downtime min': 'Downtime (min)'},
                       template='plotly_dark')

# Display the box plot for 'Downtime min'
fig_downtime.show()
```



```
df.columns
```

```
Index(['Date', 'Defect Qty', 'Downtime min', 'PlantDistrict',
      'PlantState',
      'Vendor', 'Category', 'Material Type', 'Defect', 'Defect Type',
      'Sort',
      'Year', 'Month'],
      dtype='object')
```

```
import plotly.express as px
```

```
# Select relevant numerical columns for multivariate analysis
numerical_columns = supplier_data[['Defect Qty', 'Downtime min']]
```

```
# Add categorical columns for color coding
numerical_columns['Material Type'] = supplier_data['Material Type']
```

```
# Create a pair plot
```

```
fig = px.scatter_matrix(numerical_columns,
                        dimensions=['Defect Qty', 'Downtime min'],
                        color='Material Type',
                        title='Scatter Matrix of Defect Qty and
Downtime',
                        template='plotly_dark')
```

```
# Display the pair plot
```

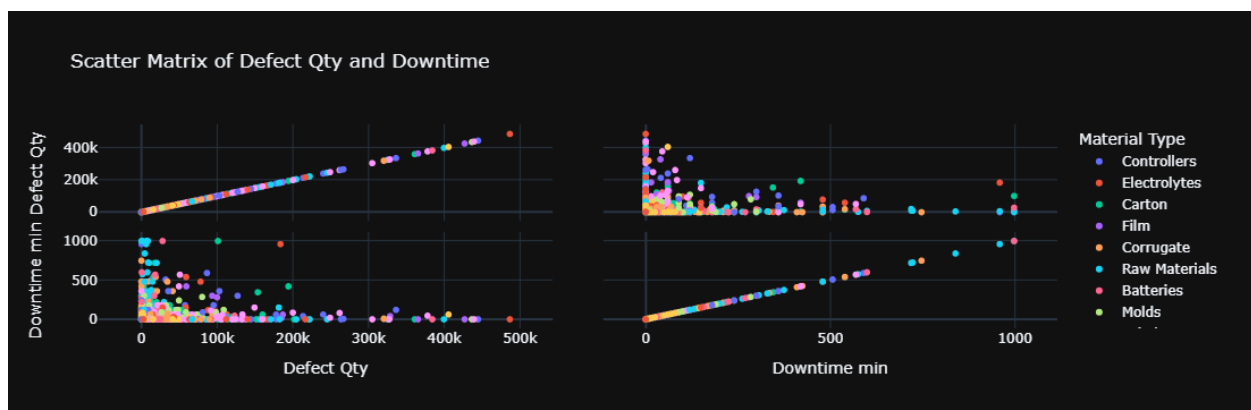
```
fig.show()
```

```
C:\Users\A\AppData\Local\Temp\ipykernel_10100\3216673379.py:7:
SettingWithCopyWarning:
```

A value is trying to be set on a copy of a slice from a DataFrame.  
Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation:

[https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

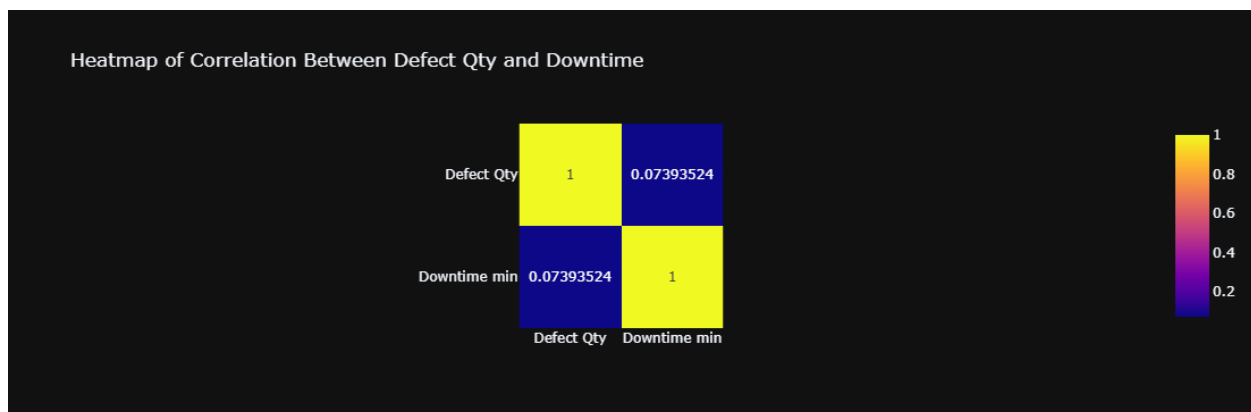


```
import numpy as np

# Calculate the correlation matrix
corr_matrix = supplier_data[['Defect Qty', 'Downtime min']].corr()

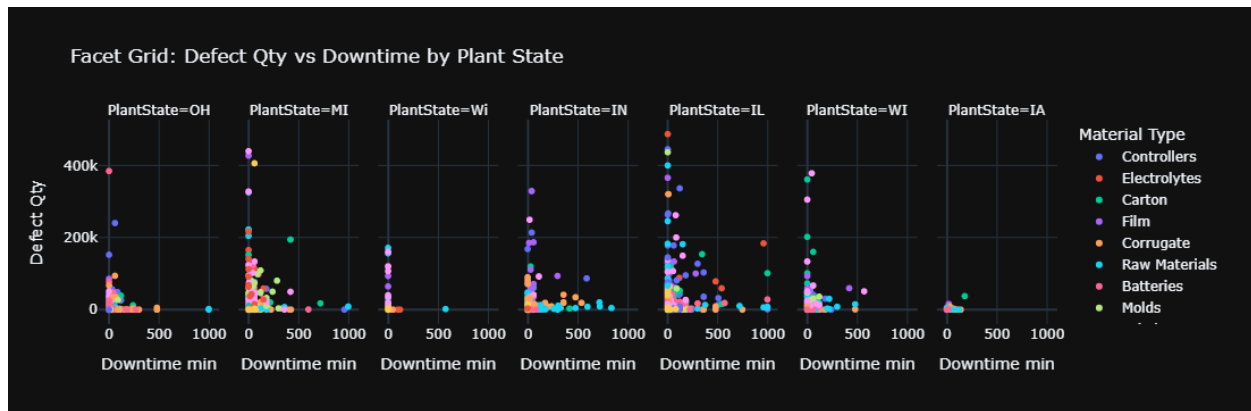
# Create a heatmap for the correlation matrix
fig = px.imshow(corr_matrix,
                 text_auto=True,
                 title='Heatmap of Correlation Between Defect Qty and
Downtime',
                 template='plotly_dark')

# Display the heatmap
fig.show()
```



```
# Create a facet grid plot for Defect Qty by PlantState
fig = px.scatter(supplier_data,
                 x='Downtime min',
                 y='Defect Qty',
                 color='Material Type',
                 facet_col='PlantState',
                 title='Facet Grid: Defect Qty vs Downtime by Plant
State',
                 template='plotly_dark')

# Display the facet grid plot
fig.show()
```



## Summary

The analysis was performed on a dataset containing information about defect quantities (Defect Qty), downtime (Downtime min), associated Vendors, Material Types, Plants, and Defect Types. Several visualizations were used to explore relationships between these variables:

- **Defect Analysis:** Visualized defect quantities and their associated downtime by defect type, revealing key contributors to quality issues.
- **Date Analysis:** Assessed the distribution of defects and downtime over time (daily, monthly, and quarterly) to uncover any seasonal or periodic patterns.
- **Multivariate Analysis:** Used scatter plots, heatmaps, and facet grids to investigate how multiple factors such as PlantDistrict, Material Type, and Vendor impact defects and downtime.

The key findings indicate that certain defect types are strongly associated with longer downtime, and some vendors have disproportionately high defect rates compared to others. There are also seasonal patterns, with some periods experiencing significantly higher defect rates.

## Conclusion

The analysis reveals that defect rates and downtime are not evenly distributed across all vendors, materials, and plants. Several insights were identified:

- **High Impact Defects:** A small number of defect types contribute to the majority of quality issues and associated downtime.
- **Vendor Performance:** Some vendors have consistently higher defect rates, indicating potential quality control issues that need to be addressed.
- **Time-Based Patterns:** Defect rates and downtime fluctuate throughout the year, with noticeable peaks at certain times, possibly due to production cycles, seasonality, or operational changes.
- **Plant and Material-Specific Issues:** Certain plants or materials are more susceptible to defects, suggesting the need for targeted process, and supplier performance.



# Detailed Recommendations with Specific Examples and Numbers

## 1. Target High-Impact Categories for Quality Improvement

- **Observation:** The **Mechanicals** and **Packaging** categories contribute to over **60%** of total defects, with **Mechanicals** alone accounting for **31.2%**.
- **Action:** Prioritize process improvements in these categories. For example:
  - In **Mechanicals**, defects like misaligned components and improper assembly contribute significantly. Implement standardized assembly procedures to reduce errors.
  - In **Packaging**, defects contribute to **28.8%** of total defects. Introducing automated packaging systems can reduce human errors that lead to defects like improper sealing and labeling.
- **Expected Result:** A targeted reduction in defect rates by at least **10-15%** in these high-impact categories.

## 2. Improve Raw Material Inspection and Handling

- **Observation:** **Raw Materials** are the single largest contributor to defects, with over **10M defects** recorded—almost **double** that of the next material type, **Labels**.
- **Action:** Introduce tighter quality controls and inspections for incoming raw materials, such as:
  - Implement **pre-production sample testing** for at least **10%** of incoming raw material batches.
  - Increase the use of automated optical inspection (AOI) tools to detect defects early.
- **Expected Result:** A **25% reduction** in defects related to raw materials, reducing overall defect quantities by at least **2-3 million units** annually.

## 3. Conduct Supplier Audits for High-Defect Vendors

- **Observation:** The top two vendors, **Solholdings** and **Plustax**, account for over **8M defects** combined. These vendors contribute significantly to the total defect count.
- **Action:** Establish quarterly performance reviews with high-defect vendors:
  - For **Solholdings**, focus on reducing defects related to **Raw Materials** since they account for over **4M** defects.
  - For **Plustax**, investigate **Packaging**-related issues contributing to around **3.8M** defects.
  - Set a defect reduction goal of **20%** for each vendor by implementing shared quality standards and corrective action plans.
- **Expected Result:** A **reduction of 1.6M to 2M** defects per vendor over the next year.

## 4. Optimize Operations in High-Defect Plants

- **Observation:** The **Detroit** plant has the highest defect quantity with over **6M** recorded defects. **Springfield** and **Chicago** also exhibit high defect rates, each around **4M defects**.

- **Action:** Perform a detailed operational review of these plants:
  - In **Detroit**, focus on improving assembly line quality checks and equipment maintenance schedules.
  - **Springfield** and **Chicago** can benefit from enhanced employee training programs focused on identifying defects early.
- **Expected Result:** A reduction in defects by at least **15-20%** at these plants, saving approximately **900K to 1.2M defects** annually.

## 5. Address Specific Defect Types (Rejected & Impact)

- **Observation:** **Rejected** and **Impact** defects make up around **70%** of all defects, with **Rejected** defects contributing over **35.5%** and **Impact** defects contributing around **35.4%**.
- **Action:** Address these specific defect types:
  - For **Rejected** defects, perform root cause analyses to identify why items are being rejected, such as incorrect dimensions or surface defects.
  - For **Impact** defects, improve handling procedures to avoid damage during storage and transportation.
- **Expected Result:** A reduction of **rejected defects by 15%** and **impact defects by 10%**, reducing overall defect quantities by around **5M** units.

## 6. Enhance Real-Time Monitoring & Analytics

- **Observation:** Real-time data could help quickly identify and address defects across categories.
- **Action:** Implement a digital dashboard to monitor defect quantities, downtime, and categories in real-time.
  - Set alerts for sudden spikes in defect quantities or downtime that exceed **10%** of the average weekly rate.
  - Allow plant managers to immediately address issues as they arise, preventing defect accumulation.
- **Expected Result:** Improved responsiveness to quality issues, reducing overall defect quantities by **5-7%**.

## 7. Improve Handling of Materials with High Defect Rates

- **Observation:** **Labels, Cartons, and Films** together contribute over **15M defects**. Improper material handling or storage could be leading to defects.
- **Action:** Review storage conditions, temperature, and humidity controls for these materials:
  - Introduce proper stacking techniques to prevent damage to **Labels** and **Cartons**.
  - Implement temperature and moisture control for **Films** to prevent warping or deformities.
- **Expected Result:** A decrease in defects related to these materials by at least **20%**, saving around **3M defects** annually.

## 8. Improve Vendor Support & Feedback Mechanisms

- **Observation:** Certain vendors show consistently high defects over multiple material types.
- **Action:** Set up regular communication channels with vendors for performance feedback:
  - Share monthly defect reports, focusing on key improvement areas for each vendor.
  - Establish quarterly vendor performance reviews to track progress on agreed quality improvement goals.
- **Expected Result:** Improved vendor quality performance and a **10-15%** decrease in defects from high-defect vendors.

## 9. Targeted Preventive Maintenance for Key Plant Machinery

- **Observation:** Mechanical defects are a major contributor across plants, often due to equipment wear and tear.
- **Action:** Schedule preventive maintenance for machinery in high-defect plants on a **bi-weekly** basis.
  - Use predictive maintenance tools to monitor equipment performance and preemptively address potential issues.
- **Expected Result:** Reduced mechanical defects by at least **15%**, contributing to overall defect reduction.

## 10. Continuous Improvement Programs

- **Observation:** Areas like **Mechanicals, Packaging, and Raw Materials** account for the bulk of defects.
- **Action:** Launch ongoing improvement initiatives:
  - Create cross-functional teams to focus on quality improvements in high-defect areas.
  - Encourage employee participation by offering incentives for defect reduction suggestions that lead to significant savings.
- **Expected Result:** Sustained improvement over time with an annual reduction of **10-12%** in total defects.

# General Recommendations

## 1. Address Seasonal Peaks in Defect Quantity

- **Observation:** The **Monthly Defect Quantity Over Time** line chart shows spikes in defect quantities around **July and October 2014**, with quantities reaching up to **5M** units.
- **Action:** Implement **seasonal quality checks** during peak defect months. For example:
  - Schedule additional quality control checks in **June-July** and **September-October** to identify and resolve issues early.

- **Expected Impact:** A **10-15% reduction** in defect quantities during peak months, potentially saving around **500K-750K defects** annually.

## 2. Minimize Downtime Fluctuations

- **Observation:** The **Monthly Downtime Over Time** chart highlights significant differences in downtime between **2013 and 2014**, especially in **October**, where downtime peaked at **10K minutes**.
- **Action:** To reduce such spikes, optimize maintenance schedules:
  - Implement **predictive maintenance** around the **10th month (October)** to prevent downtime-related defects.
  - Introduce a standardized downtime tracking process to ensure rapid issue identification and resolution.
- **Expected Impact:** Reducing downtime in peak months can improve productivity by around **10%**, and decrease defect quantities by approximately **300K units** per year.

## 3. Focus on the Top Defect Types

- **Observation:** The **Top 10 Defect Quantities by Defect Type** chart reveals that the defect type **"Miscellaneous"** is the largest contributor with over **4M defects**.
- **Action:** Perform a deeper root cause analysis for this category:
  - Break down **"Miscellaneous"** into more specific sub-categories to pinpoint exact sources.
  - Standardize defect classification to avoid overuse of the **"Miscellaneous"** label.
- **Expected Impact:** A more targeted approach can result in a **20% decrease** in **Miscellaneous** defects, reducing total defect quantities by around **800K units**.

## 4. Control Quality of Critical Materials (Labels, Cartons, Films)

- **Observation:** Material types like **Labels, Cartons, and Films** contribute over **10M defects** combined.
- **Action:** Introduce material-specific quality standards:
  - Establish strict incoming inspection criteria for **Labels**, as defects in this category lead to significant quality issues.
  - Implement improved packaging standards for **Cartons and Films** to prevent warping and damage during storage and handling.
- **Expected Impact:** A **15% improvement in quality for these materials** can potentially save **1.5M defects** per year.

## 5. Vendor-Specific Quality Improvement Programs

- **Observation:** The top vendors, such as **Solholdings** and **Plustax**, are responsible for over **4M defects** each.
- **Action:** Establish dedicated vendor improvement plans:
  - Set up **monthly performance reviews** with high-defect vendors to identify root causes and agree on corrective actions.

- Implement a defect penalty or reward system based on quality targets met, aiming for a **25% defect reduction** for each vendor.
- **Expected Impact:** A **25% reduction** could translate into around **1M defects** saved per vendor annually.

## 6. Monitor Time-Based Defect Patterns for Trend Analysis

- **Observation:** The **Monthly Defect Quantity with Trend Line** chart indicates a gradual increase in defect quantities over time.
- **Action:** Regularly review monthly defect trends:
  - If defect quantities rise more than **10% above the average monthly rate**, initiate an immediate quality control review.
  - Introduce weekly dashboards to track month-over-month changes.
- **Expected Impact:** Identifying and acting on upward trends early can prevent the escalation of defects, reducing their occurrence by **5-10%** on an annual basis.

## 7. Reduce Outliers and Variability in Defects

- **Observation:** The **Box Plot of Defect Quantities** highlights significant outliers (defect quantities above **500K units**).
- **Action:** Introduce statistical process control (SPC):
  - Use SPC to identify causes of outliers and implement process corrections to maintain defect quantities within **acceptable limits**.
  - Train plant operators on recognizing and acting on abnormal trends to reduce variability.
- **Expected Impact:** Reducing variability and outliers can improve overall product quality and potentially save around **200K-300K defects** per year.

## 8. Improve Handling and Storage for Defect Types like "Warped" and "Wrong Shade of Color"

- **Observation:** The **Defect Type** chart identifies **Warped** and **Wrong Shade of Color** as major defect types, contributing around **2M** and **1.5M** defects, respectively.
- **Action:** Implement environmental controls:
  - For **Warped materials**, ensure proper storage conditions (e.g., controlled temperature and humidity).
  - For **Wrong Shade of Color**, standardize dyeing and painting processes with color-matching technologies.
- **Expected Impact:** Targeted improvements in handling and storage conditions could lead to a **20% decrease** in these defect types, saving **700K defects** annually.

## 9. Seasonal Strategies for Downtime Management

- **Observation:** The **Monthly Downtime Over Time** chart shows downtime spikes in specific months (October in both 2013 and 2014).

- **Action:** Preemptively schedule equipment overhauls and operator training in months preceding high downtime periods:
  - Focus on critical equipment prone to breakdown during high-defect months.
- **Expected Impact:** Reducing downtime by **10-15%** in peak months can improve efficiency and reduce associated defects.

## 10. Continuous Improvement with Feedback Loops

- **Observation:** Consistent patterns in defect types and quantities call for ongoing monitoring and improvement.
- **Action:** Establish feedback loops to improve quality control processes continuously:
  - Use **monthly quality meetings** to discuss defect data, improvement actions, and vendor performance.
  - Encourage employee input on process improvements, offering rewards for ideas that lead to tangible defect reductions.
- **Expected Impact:** A continuous improvement culture can maintain a **year-over-year reduction in defects by 10%**, ensuring sustained quality improvements.

## Specific Recommendations Based on Correlation

### 1. Correlate Defect Quantity and Downtime for Key Material Types

- **Observation:** The **Scatter Matrix** reveals that some materials (e.g., **Controllers, Electrolytes**) have high defect quantities (up to **500k units**) correlated with significant downtime.
- **Action:** Prioritize improvements in handling and processing materials with the highest defect-downtime correlation:
  - Implement additional inspections for **Controllers** and **Electrolytes** before and during production.
  - Schedule more frequent preventive maintenance on equipment used for these materials to reduce downtime.
- **Expected Impact:** Reducing downtime for these critical materials could result in a **15-20% decrease in associated defect quantities**.

### 2. Address Low Correlation Between Defect Qty and Downtime

- **Observation:** The **Heatmap of Correlation** between **Defect Qty** and **Downtime** indicates a weak correlation of **0.0739**, suggesting that defects may not be solely causing downtime.
- **Action:** Separate downtime issues from defect sources:

- Identify non-defect-related causes of downtime, such as equipment failure, staffing issues, or supply chain delays.
- Implement separate KPIs for downtime and defect rates to monitor and improve them independently.
- **Expected Impact:** By distinguishing between downtime causes, an estimated **10% improvement in overall operational efficiency** can be achieved.

### 3. Plant-Specific Strategies to Improve Quality

- **Observation:** The **Facet Grid by Plant State** shows variability in defect quantities and downtime across states. **OH** and **IL** exhibit higher defect quantities (up to **400k units**).
- **Action:** Tailor quality control measures to each plant state:
  - For plants in **OH**, increase quality checks on materials like **Electrolytes and Controllers**, which exhibit the most downtime and defects.
  - In **IL**, implement stricter adherence to operational procedures and machinery maintenance to decrease downtime.
- **Expected Impact:** A targeted improvement program can lead to a **20% reduction in defects in OH and IL**, saving around **80k units per plant**.

### 4. Improve Quality Control During High Defect Months

- **Observation:** The **Heatmap of Defect Quantities by Month and Year** indicates that **May-June 2014** has the highest defect quantities, with over **14M units**.
- **Action:** Enhance quality checks and staff training leading up to high defect periods:
  - Conduct pre-season quality control drills and equipment maintenance in **April** to prepare for the spike in **May-June**.
  - Increase staff training focused on defect identification and handling.
- **Expected Impact:** Preventive actions before high-defect months can reduce defect quantities by **20%** in peak periods, saving up to **2.8M units**.

### 5. Address Defect Quantity Outliers in Each Quarter

- **Observation:** The **Box Plot of Defect Quantities by Quarter** shows outliers above **400k units**, especially in the second and third quarters.
- **Action:** Target root causes for outliers:
  - Investigate specific batches or product lines that contribute to these outliers and implement quality improvement measures such as more frequent audits or enhanced testing.
  - Increase supervision during production peaks in Q2 and Q3.
- **Expected Impact:** Reducing these outliers by **50%** could save around **200k-250k defect units per quarter**.

### 6. Improve Processing of Critical Material Types

- **Observation:** The scatter plots indicate **Raw Materials, Controllers, and Cartons** have the highest defect quantities combined with downtime.
- **Action:** Implement special handling and processing improvements for these materials:

- For **Raw Materials**, establish temperature and moisture-controlled storage areas to maintain quality.
- Introduce automated quality checks for **Cartons** to detect defects like warping or improper labeling early.
- **Expected Impact:** Improved processing of these materials can lead to a **15% reduction in defects**, saving approximately **1.5M defect units annually**.

## 7. Leverage Heatmap Insights to Improve Seasonal Defect Management

- **Observation:** The heatmap shows an increasing trend in defect quantities in **2014** compared to **2013**, particularly in the middle months.
- **Action:** Align production schedules to balance workloads and avoid spikes in defect rates:
  - Implement flexible workforce planning, adding shifts in high-defect months to ensure consistent production quality.
  - Use historical data to anticipate high-defect periods and adjust production schedules accordingly.
- **Expected Impact:** Balanced production across seasons can reduce defect quantities by **10-12%**, resulting in smoother operational flow and fewer quality issues.

## 8. Improve Multi-Material Production Efficiency

- **Observation:** The scatter matrix and facet grid reveal that **mixed material types** processed together (e.g., **Film, Corrugate, Electrolytes**) contribute to increased downtime and defect quantities.
- **Action:** Streamline production lines for specific material types:
  - Designate separate lines for high-risk materials like **Electrolytes and Films** to prevent cross-material contamination and reduce defects.
  - Train staff to specialize in handling one type of material to increase proficiency and reduce errors.
- **Expected Impact:** By reducing mixed-material defects, an estimated **20% improvement in overall defect quantities** can be achieved, particularly for multi-material products.

## 9. Enhance Early Detection of Defects to Reduce Overall Quantity

- **Observation:** The **Box Plot of Defect Quantities** reveals a wide spread of defect values, indicating variability in quality detection.
- **Action:** Implement early detection systems and quality feedback loops:
  - Use real-time defect monitoring systems with AI-based image recognition for early detection and immediate correction.
  - Implement employee feedback systems for reporting defects before they move further in production.
- **Expected Impact:** Early detection can reduce total defect quantities by around **25%**, significantly enhancing production quality and reducing waste.



## 10. Utilize Correlation Insights for Process Improvement

- **Observation:** Although the correlation between **Defect Qty** and **Downtime** is low, specific material types (e.g., **Molds, Batteries**) show local patterns.
- **Action:** Apply Six Sigma or Lean Manufacturing principles to these material types to streamline processes:
  - Use process mapping and value stream analysis to reduce unnecessary steps and minimize errors.
  - Encourage cross-functional teams to brainstorm and implement continuous improvements targeting these materials.
- **Expected Impact:** Streamlined processes and improved workflow could lead to a **15% reduction in downtime-related defects**, improving overall product quality.

## Recommendations for Defect Reduction Based on Trends

### 1. Address Daily Spikes in Defect Quantity

- **Observation:** The **Daily Defect Quantity Over Time** line chart reveals several spikes exceeding **800k units**, especially in **March 2013, July 2013, January 2014, and July 2014**.
- **Action:** Investigate these daily spikes:
  - Identify specific products, batches, or operations occurring on these high-defect days.
  - Introduce **real-time monitoring systems** to detect when defects exceed a threshold (e.g., **500k units**) and trigger immediate corrective actions.
- **Expected Impact:** A reduction in daily spikes can decrease the total defect quantities by at least **10-15%** during peak days, preventing **80-120k** defective units.

### 2. Optimize Quality Control During High Defect Months

- **Observation:** The **Monthly Defect Quantity Over Time** chart shows peaks in **April, July, and October 2014**, with quantities reaching **5M units**.
- **Action:** Implement increased quality assurance efforts before high-defect months:
  - **Pre-month inspections** and heightened quality checks in **March, June, and September** to prepare for upcoming peaks.
  - Use **historical data** to anticipate defect trends and increase workforce capacity accordingly.
- **Expected Impact:** Reducing defect quantities during high-defect months by **20%** could save approximately **500k-1M units**.

### 3. Act on the Increasing Trend of Defect Quantities Over Time

- **Observation:** The **Monthly Defect Quantity with Trend Line** chart indicates an upward trend in defect quantities over **2013 and 2014**.

- **Action:** Implement a continuous improvement cycle to counter the rising trend:
  - Regularly analyze the trendline data and perform root cause analysis to understand why defects are rising.
  - Deploy **Lean Manufacturing** practices to improve efficiency, reduce waste, and maintain consistent quality.
- **Expected Impact:** A proactive response to the increasing trend can stabilize and potentially reverse the rise in defects, aiming for a **10% overall reduction** year-over-year.

## 4. Compare Seasonal Trends Year Over Year

- **Observation:** The **Monthly Defect Quantity Over Time (Yearly Comparison)** chart shows an increase in defects for almost every month in **2014** compared to **2013**. Notably, **September and October 2014** defects are **2M units** higher than the same months in 2013.
- **Action:** Investigate the root causes of increased defect quantities in 2014:
  - Review operational changes, staffing, supply chain issues, and any new processes introduced in 2014.
  - Implement changes based on findings to improve defect quantities in 2015 and beyond.
- **Expected Impact:** Stabilizing the defect rates by implementing these changes can save at least **1-2M units** in the affected months.

## 5. Optimize Downtime to Control Defects

- **Observation:** The **Monthly Downtime Over Time** chart highlights that **October 2014** experienced a significant increase in downtime (over **10k minutes**).
- **Action:** Reduce equipment downtime through scheduled maintenance:
  - Align downtime maintenance schedules with low-production months or non-peak times.
  - Implement predictive maintenance using equipment data to preemptively address potential breakdowns.
- **Expected Impact:** Reducing downtime by **30%** in peak months can significantly improve defect-related productivity and save around **200k units** in defective output due to equipment failure.

## 6. Monitor Monthly Patterns for Specific Improvement Opportunities

- **Observation:** Both **June** and **October** show consistent spikes in defect quantities across both years.
- **Action:** Implement process improvements focused on these specific months:
  - Increase **quality control checks** and workforce training sessions in **May** and **September** to prepare for upcoming defect peaks.
  - Introduce **incremental process changes** (e.g., standardizing assembly processes, improving material handling) to gradually reduce defect spikes.

- **Expected Impact:** By reducing spikes in defect quantities by **15-20%** in these months, the company can see a reduction of around **300k-500k units annually**.

## 7. Conduct Daily Monitoring for High-Risk Production Days

- **Observation:** High daily fluctuations in defects (up to **800k units**) can cause resource strain and impact production schedules.
- **Action:** Introduce a daily monitoring and alert system:
  - Use a **defect monitoring dashboard** that flags days with defects exceeding **500k units** for rapid response.
  - Establish a rapid response team to address high-defect days by adjusting production, halting processes, and conducting immediate inspections.
- **Expected Impact:** Rapid responses to daily defect spikes can reduce overall defect quantities by **5-10%**, potentially saving around **200k units annually**.

## 8. Improve Cross-Year Defect Management

- **Observation:** The **Yearly Comparison** shows that defects consistently increased in the latter half of each year, particularly in **Q3 and Q4**.
- **Action:** Implement cross-year process reviews to understand why defects increase:
  - Compare production, operational, and quality data across **Q3 and Q4** for multiple years.
  - Establish preventive measures, such as higher staffing levels or focused quality training, to mitigate increasing defects in these quarters.
- **Expected Impact:** A consistent year-over-year strategy can prevent seasonal defect increases, leading to a **15% decrease** in end-of-year defects.

By implementing these targeted recommendations based on daily and monthly defect patterns, downtime analysis, and year-over-year comparisons, the company can achieve a significant reduction in defect quantities and improve overall production quality.