

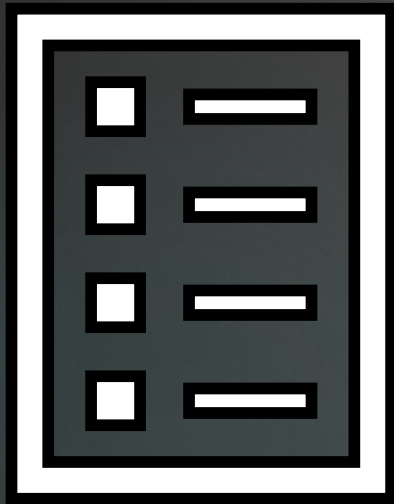


SALARY ANALYSIS REPORTS

FOR SAN FRANCISCO EMPLOYEES

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Objective

- The project aims to analyze employee compensation data, including Base Pay, Overtime Pay, Other Pay, Benefits, and their relation to Total Pay and Total Pay Benefits. This is achieved through **Exploratory Data Analysis (EDA)** and **Visualization** using **Python** in Jupyter Notebook and **Microsoft Power BI Desktop**.

Data Overview

Columns in the dataset:

- **EmployeeName**: Name of the employee.
- **JobTitle**: Title of the job.
- **BasePay**: Base salary pay.
- **OvertimePay**: Pay for overtime work.
- **OtherPay**: Any other types of compensation.
- **Benefits**: Benefits provided to the employee.
- **TotalPay**: The total pay without benefits.
- **TotalPayBenefits**: Total pay with benefits included.
- **Year**: The year of the payroll record.

Data Types

```
In [5]: df.dtypes
```

```
Out[5]: EmployeeName    object
        JobTitle        object
        BasePay         object
        OvertimePay     object
        OtherPay        object
        Benefits        float64
        TotalPay        float64
        TotalPayBenefits int64
        Year            object
        dtype: object
```

Tools & Technologies Used



Data Cleaning

Handling Null Values

```
In [13]: df.isnull().sum().sum()
```

```
Out[13]: 608
```

```
In [8]: # Converting required columns from 'object' to 'numeric' data type
df['BasePay'] = pd.to_numeric(df['BasePay'], errors = 'coerce')
df['OvertimePay'] = pd.to_numeric(df['OvertimePay'], errors = 'coerce')
df['OtherPay'] = pd.to_numeric(df['OtherPay'], errors = 'coerce')
```

```
In [15]: df['BasePay'].fillna(df['BasePay'].mean(), inplace=True)
```

```
In [17]: df=df.dropna(subset=['OvertimePay', 'OtherPay'])
```

```
In [19]: df.isnull().sum().sum()
```

```
Out[19]: 0
```

Typecasting

```
In [10]: df.dtypes
```

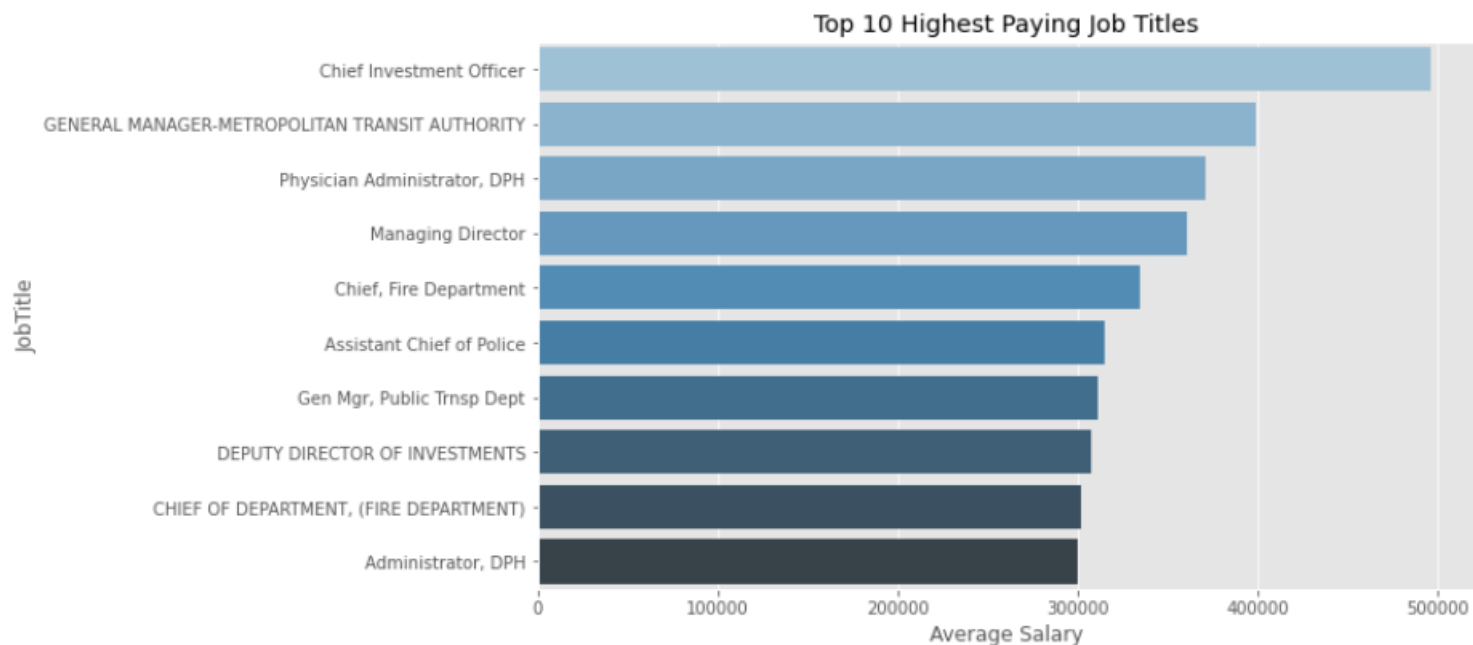
```
Out[10]: EmployeeName      object
JobTitle      object
BasePay      float64
OvertimePay    float64
OtherPay      float64
Benefits      object
TotalPay      float64
TotalPayBenefits float64
Year          int64
dtype: object
```


Exploratory Data Analysis

Highest paing Job Titles

```
In [27]: # Group by job title and get the mean TotalPay
job_salary=df.groupby('JobTitle')['TotalPay'].mean().sort_values(ascending=False).head(10)

# Plot
plt.figure(figsize=(10,6))
sns.barplot(x=job_salary.values, y=job_salary.index, palette='Blues_d')
plt.title('Top 10 Highest Paying Job Titles')
plt.xlabel('Average Salary')
plt.show()
```



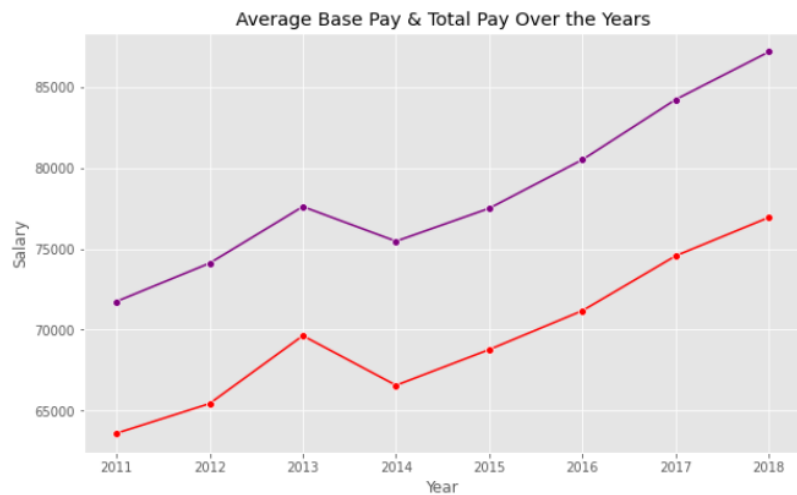
Line Charts

Line Chart Showing

1. Average Pay Over Years

```
In [28]: # Group by Year and calculate mean total pay
pay_over_years = df.groupby('Year')['TotalPay'].mean()
basepay_over_years = df.groupby('Year')['BasePay'].mean()

# Plot
plt.figure(figsize=(10,6))
sns.lineplot(x=pay_over_years.index, y=pay_over_years.values, marker='o', color='purple')
sns.lineplot(x=basepay_over_years.index, y=basepay_over_years.values, marker='o', color='red')
plt.title('Average Base Pay & Total Pay Over the Years')
plt.xlabel('Year')
plt.ylabel('Salary')
plt.show()
```



2. Number of employees and jobs over the Years

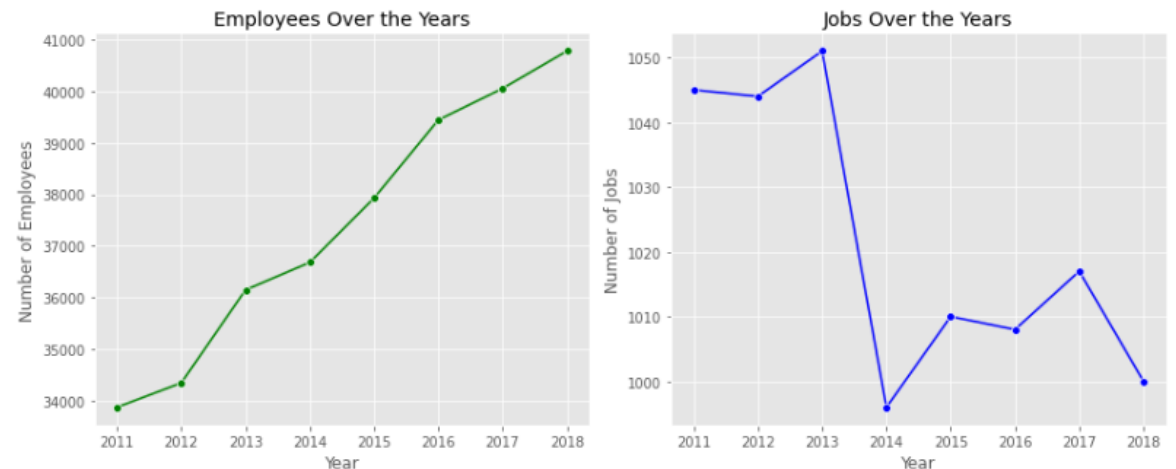
```
In [29]: # Calculate number of employees and jobs group by Year
employee_over_years = df.groupby('Year')['EmployeeName'].nunique()
job_over_years = df.groupby('Year')['JobTitle'].nunique()

# Plot Linechart for Number of employees and jobs over the Years
plt.figure(figsize=(12,5))

plt.subplot(1,2,1)
sns.lineplot(x=employee_over_years.index, y=employee_over_years.values, marker='o', color='green')
plt.title('Employees Over the Years')
plt.xlabel('Year')
plt.ylabel('Number of Employees')

plt.subplot(1,2,2)
sns.lineplot(x=job_over_years.index, y=job_over_years.values, marker='o', color='blue')
plt.title('Jobs Over the Years')
plt.xlabel('Year')
plt.ylabel('Number of Jobs')

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



Bar Charts

```
In [30]: # Plot barcharts for BasePay, TotalPay, TotalPayBenefits, OvertimePay, and OtherPay
plt.figure(figsize=(12,13))

plt.subplot(3,2,1)
sns.barplot(data=df, x='Year', y='BasePay')
plt.xticks(rotation='vertical')
plt.title('BasePay over Years')

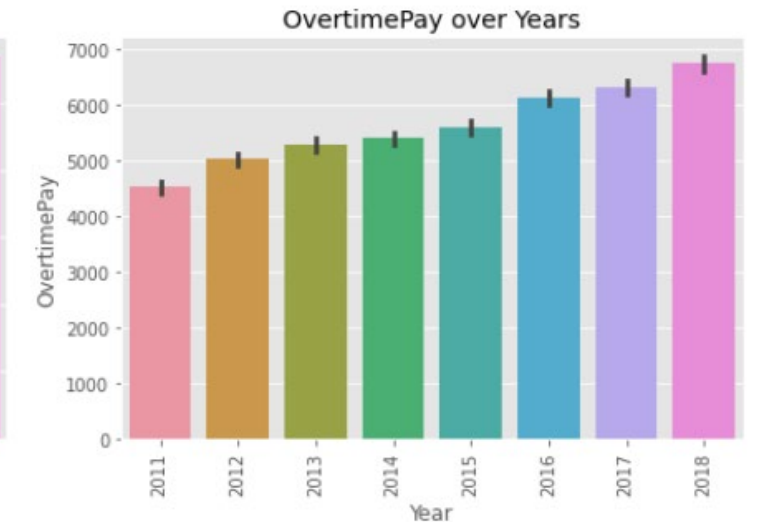
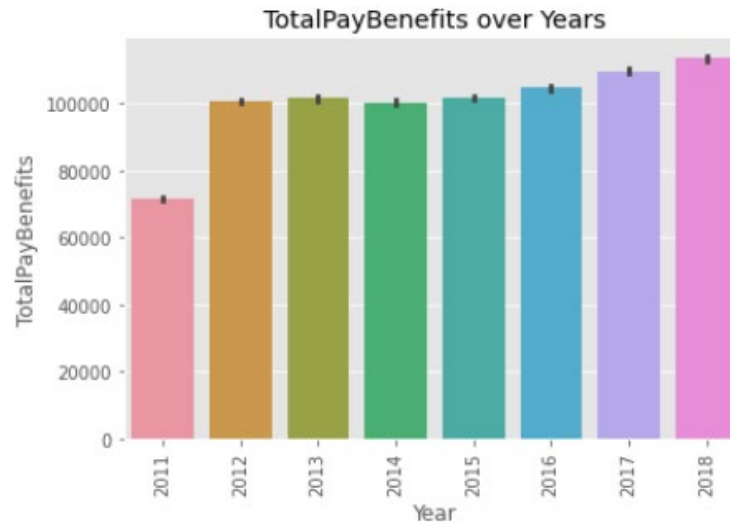
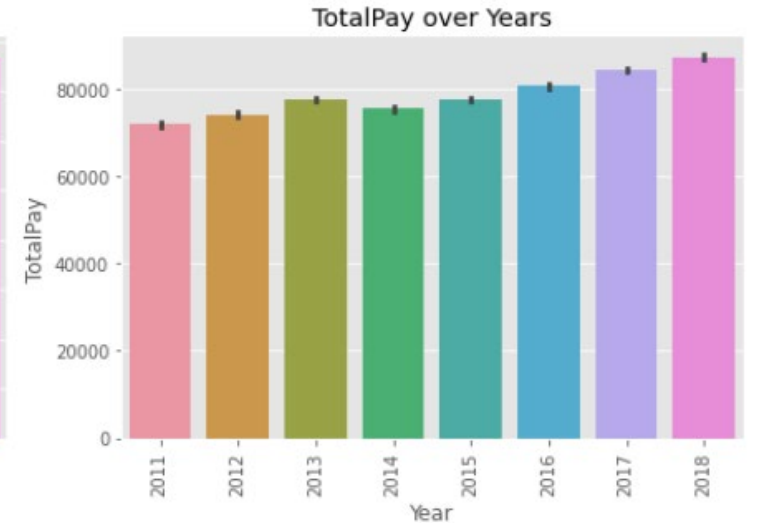
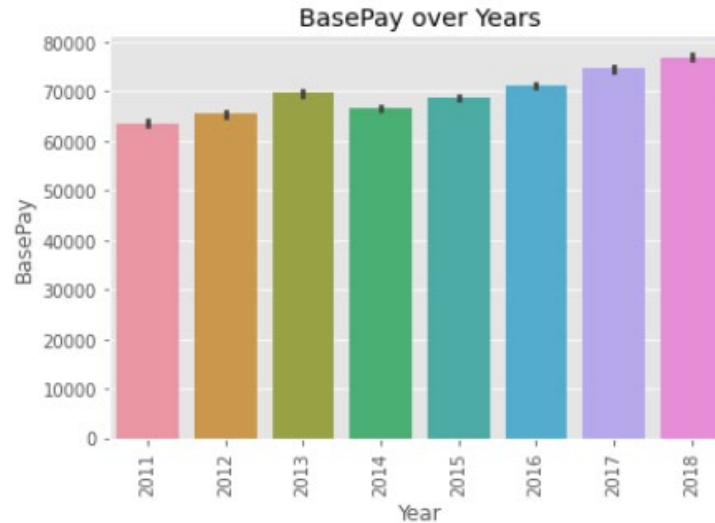
plt.subplot(3,2,2)
sns.barplot(data=df, x='Year', y='TotalPay')
plt.xticks(rotation='vertical')
plt.title('TotalPay over Years')

plt.subplot(3,2,3)
sns.barplot(data=df, x='Year', y='TotalPayBenefits')
plt.xticks(rotation='vertical')
plt.title('TotalPayBenefits over Years')

plt.subplot(3,2,4)
sns.barplot(data=df, x='Year', y='OvertimePay')
plt.xticks(rotation='vertical')
plt.title('OvertimePay over Years')

plt.subplot(3,2,5)
sns.barplot(data=df, x='Year', y='OtherPay')
plt.xticks(rotation='vertical')
plt.title('OtherPay over Years')

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



Graphs Showing distribution of payments

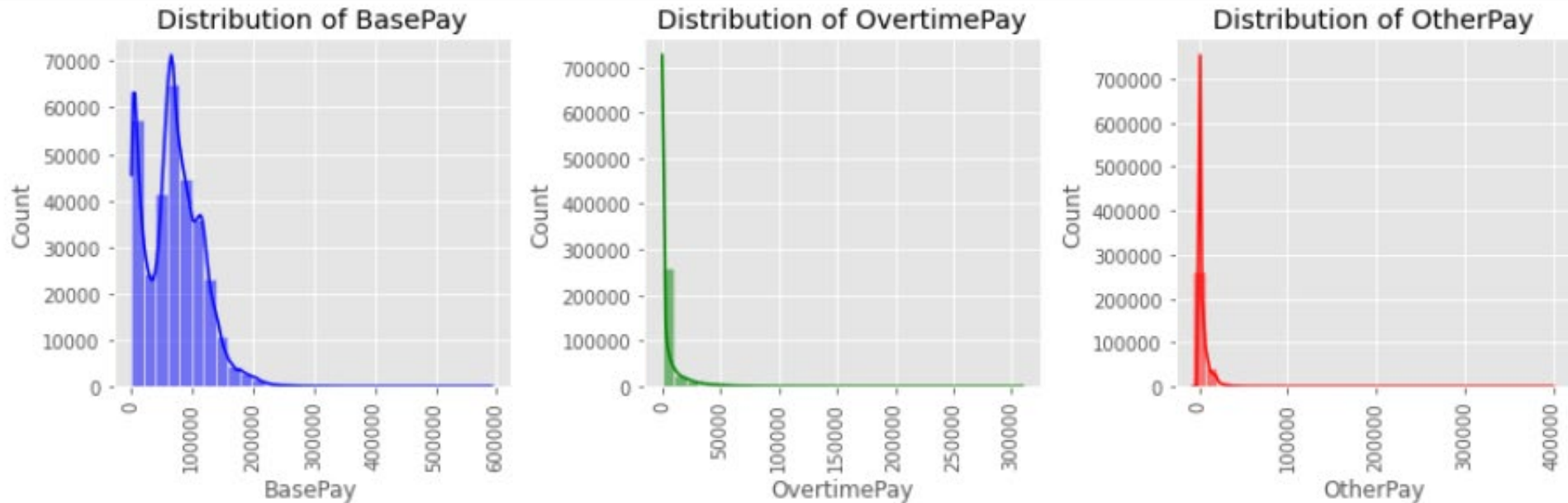
```
In [31]: # Plot histograms for BasePay, OvertimePay, and OtherPay
plt.figure(figsize=(12,4))

plt.subplot(1,3,1)
sns.histplot(df['BasePay'], bins=30, kde=True, color='blue')
plt.xticks(rotation='vertical')
plt.title('Distribution of BasePay')

plt.subplot(1,3,2)
sns.histplot(df['OvertimePay'], bins=30, kde=True, color='green')
plt.xticks(rotation='vertical')
plt.title('Distribution of OvertimePay')

plt.subplot(1,3,3)
sns.histplot(df['OtherPay'], bins=30, kde=True, color='red')
plt.xticks(rotation='vertical')
plt.title('Distribution of OtherPay')

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



Pie Charts

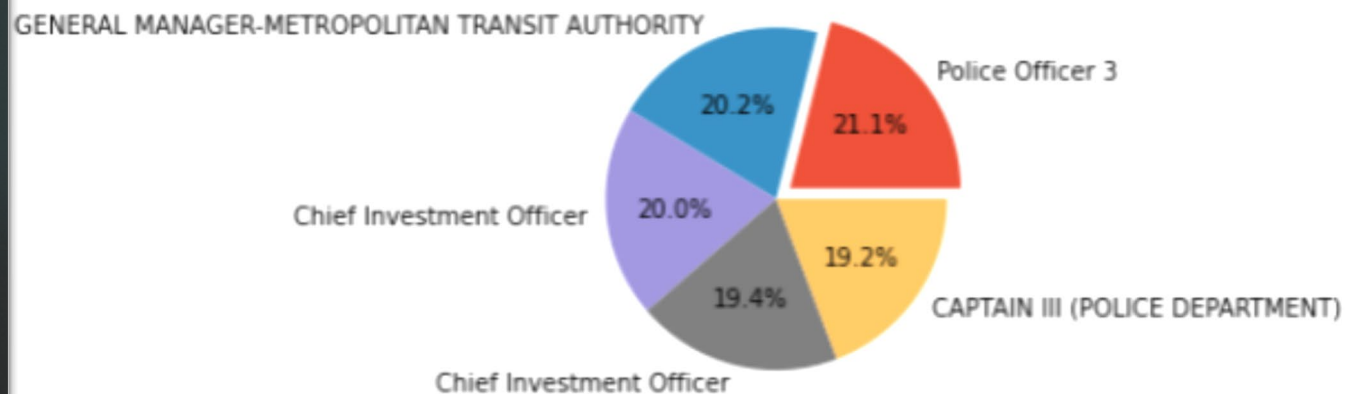
```
In [45]: plt.figure(figsize=(15,3))

plt.subplot(1,2,1)
tmp = df.sort_values(by=['TotalPay'], ascending = False)
tmp = tmp[['JobTitle', 'TotalPay']].head()
plt.pie(tmp.TotalPay, labels = tmp.JobTitle, autopct='%1.1f%%', explode=[0.1,0,0,0,0])
plt.title('Total Pay by Job Title')

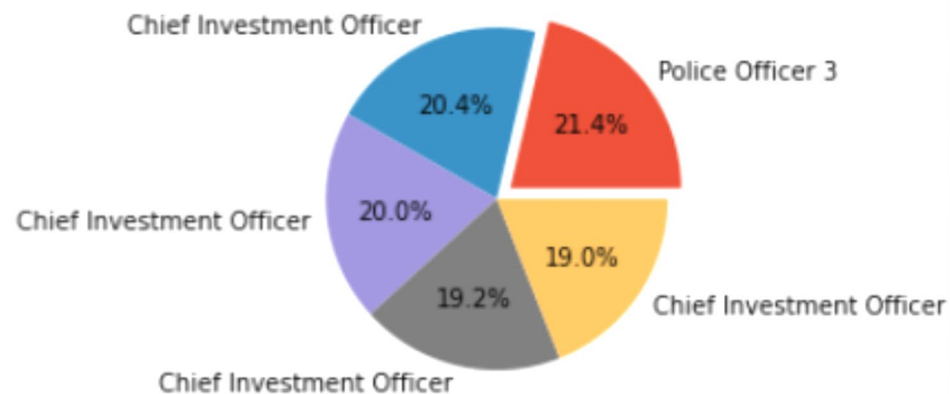
plt.subplot(1,2,2)
tmp = df.sort_values(by=['TotalPayBenefits'], ascending = False)
tmp = tmp[['JobTitle', 'TotalPayBenefits']].head()
plt.pie(tmp.TotalPayBenefits, labels = tmp.JobTitle, autopct='%1.1f%%', explode=[0.1,0,0,0,0])
plt.title('Total Pay Benefits by Job Title')

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

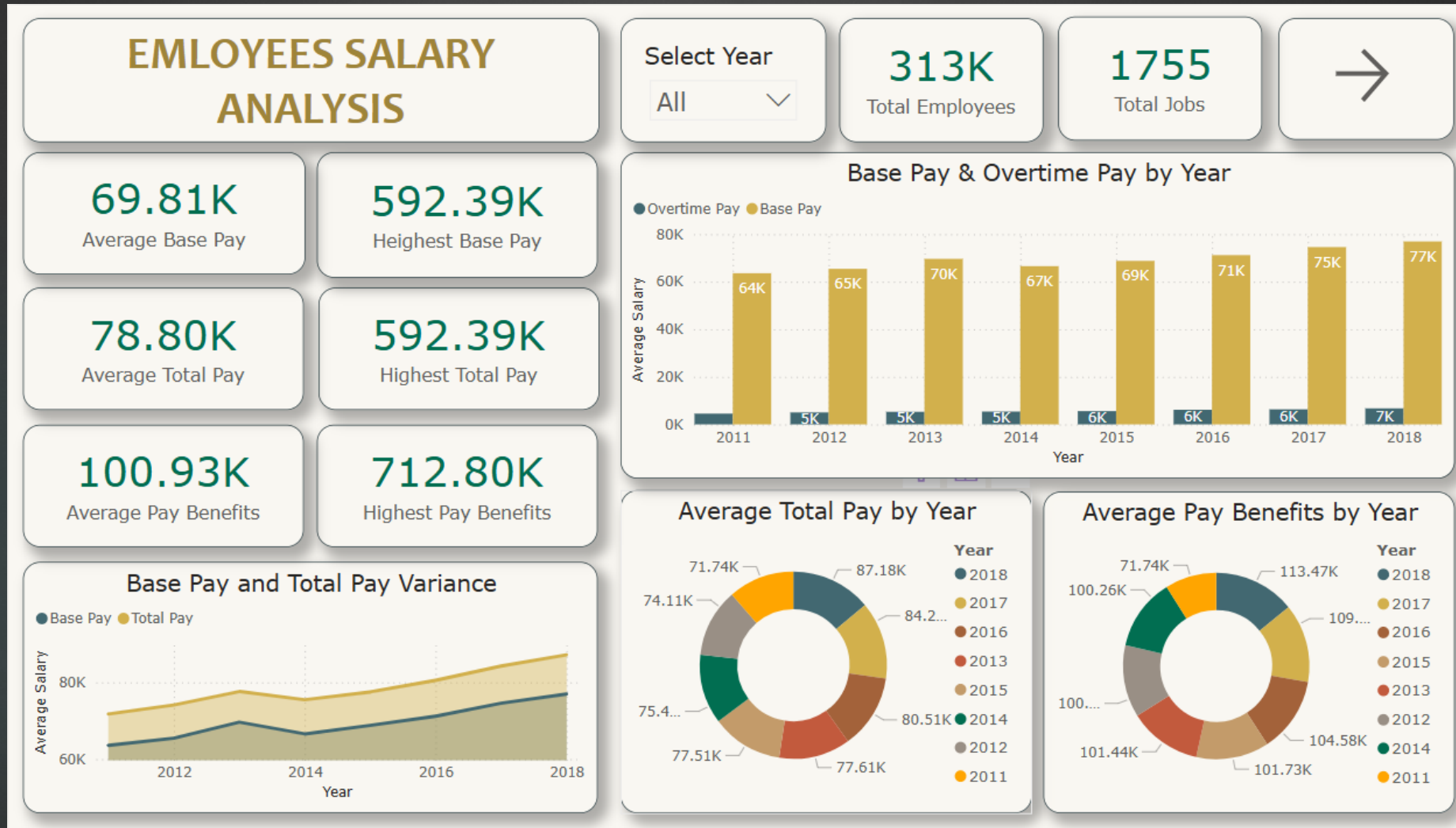
Total Pay by Job Title



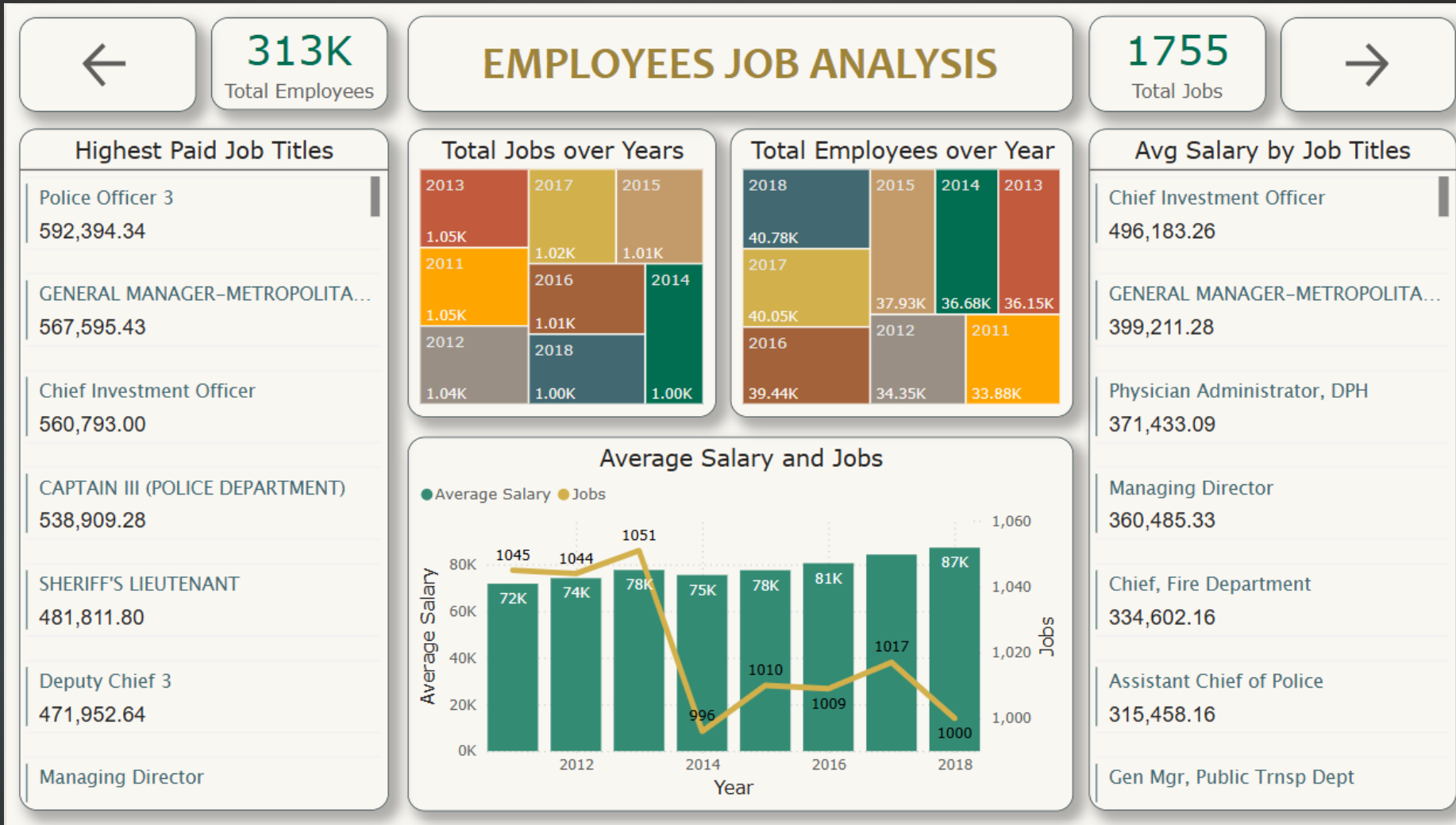
Total Pay Benefits by Job Title



Dashboard Components



Dashboard Components

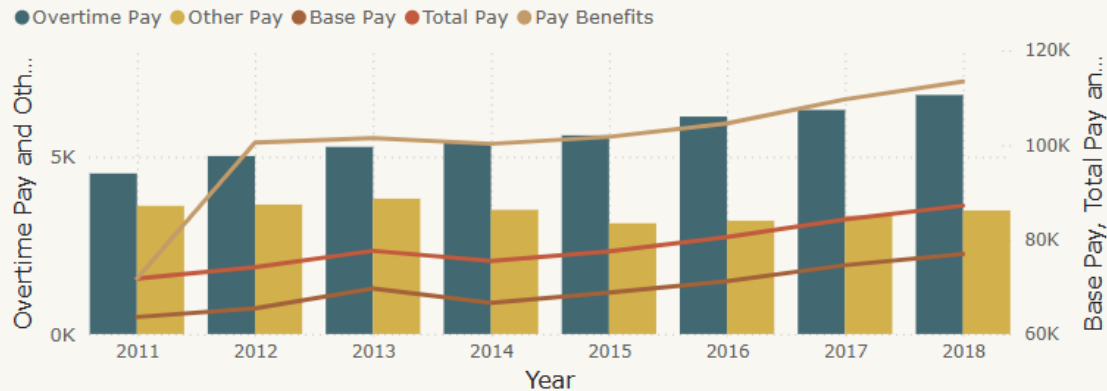


Dashboard Components

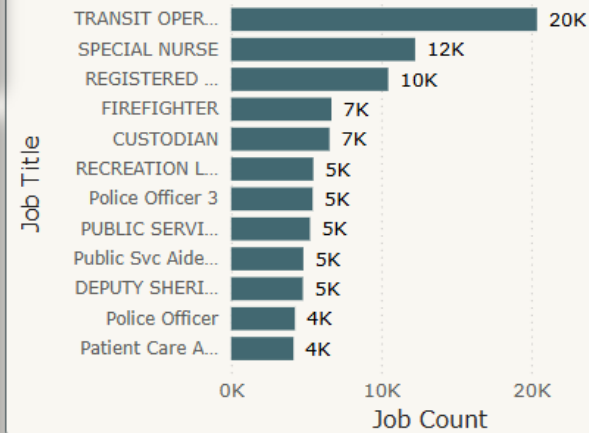


FURTHER ANALYSIS

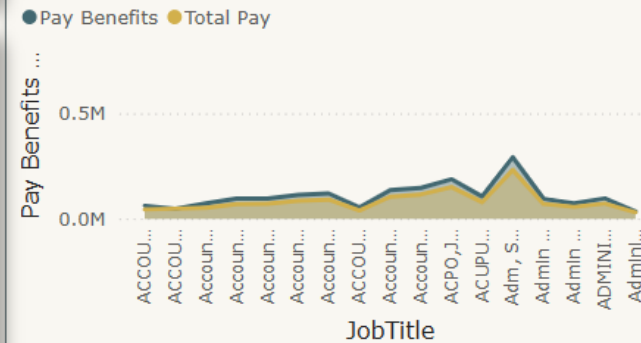
Variance of different Payments by Year



Total Jobs by Job Title



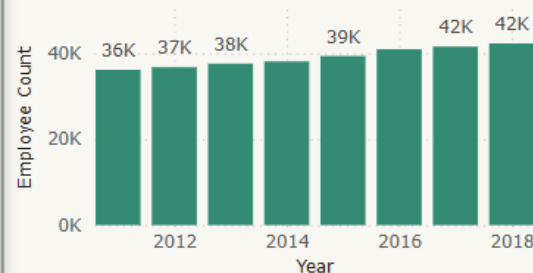
Avg Pay Benefits & Avg Salary by Job Titles



Avg Pay Benefits & Salary by Year



Total Employees by Year



Key Metrics Summary

Salary Overview:

- ▶ **Average Base Pay:** \$69.81K
- ▶ **Highest Base Pay:** \$592.39K
- ▶ **Average Total Pay:** \$78.80K
- ▶ **Highest Total Pay:** \$592.39K
- ▶ **Average Pay Benefits:** \$100.93K
- ▶ **Highest Pay Benefits:** \$712.80K

Workforce Overview:

- ▶ **Total Employees:** 313K
- ▶ **Total Job Titles:** 1,755

Trend Analysis

Base Pay & Overtime Trends

- ▶ From 2011 to 2018, **Base Pay** consistently increased from \$64K to \$77K.
- ▶ **Overtime Pay** also showed a gradual increase, reflecting growing demand or workload.

Year-wise Variance

- ▶ The line and bar charts show a **steady upward trend in all salary components**, especially after 2014.
- ▶ Benefits and overtime form a growing share of total pay, suggesting either better benefit programs or greater reliance on overtime.

Total Pay & Pay Benefits

- ▶ **Total Pay** rose from \$72K (2011) to \$87K (2018).
- ▶ **Pay Benefits** climbed significantly from \$72K to \$113K, indicating a shift toward benefits-focused compensation.

Highest Paid Titles

- ▶ **Police Officer 3:** \$592,394
- ▶ **General Manager–Metropolitan:** \$567,595
- ▶ **Chief Investment Officer:** \$560,793
- ▶ **Captain III (Police Department):** \$538,909
- ▶ High earners are predominantly from **law enforcement and executive roles**.

Trend Analysis

Average Salary by Title

- ▶ **Chief Investment Officer:** \$496,183 (highest average)
- ▶ **Assistant Chief of Police:** \$315,458
- ▶ Reflects that **administrative and public safety roles** dominate top average salaries.

Total Jobs by Year

- ▶ Job counts have remained relatively stable (~1,000 per year).
- ▶ Minor fluctuations suggest consistent hiring practices or job reclassifications.

Employee Count Over Years

- ▶ Number of employees increased from **~34K in 2011** to **~42K in 2018**.
- ▶ Growth implies workforce expansion or regular onboarding.

Jobs by Title

- ▶ **Transit Operator:** 20K jobs
- ▶ **Special Nurse:** 12K jobs
- ▶ **Registered Nurse:** 10K jobs
- ▶ Job volume indicates **healthcare and transit are the most populous sectors**.

Trend Analysis

Additional Insights:

Average Pay Benefits vs Avg Salary by Job

- ▶ The chart shows job titles with **disproportionately higher benefits** compared to salary.
- ▶ Indicates potential areas of **cost optimization or restructuring**.

Component Variance

- ▶ From 2011 to 2018, **overtime, pay benefits, and base pay** showed consistent growth.
- ▶ Increasing **benefit load** suggests a strategic shift or legal compliance changes.

Conclusion

- This analysis provides a detailed view of the compensation landscape across years and job titles. With rising salaries, expanding benefits, and consistent job volumes, the data supports informed decision-making for HR planning, salary benchmarking, and organizational budgeting.



THANK YOU