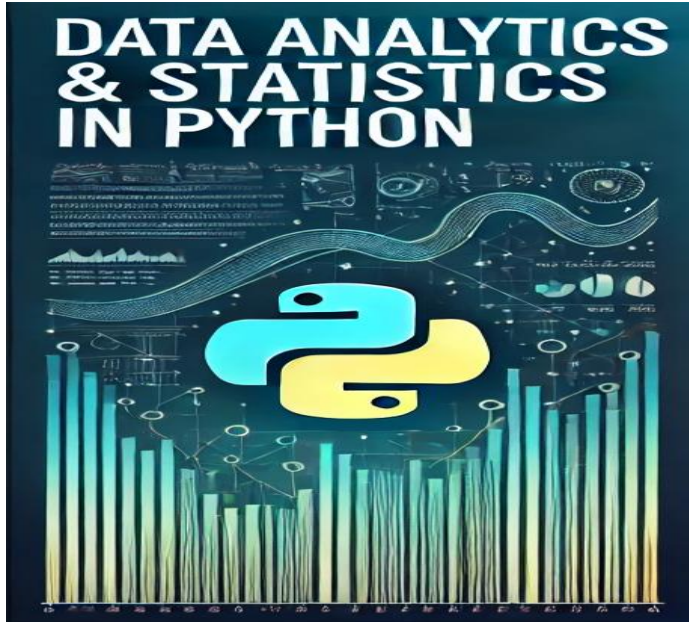


Data Analytics & Statistics in Python

Good & Bad Visualisation



Learning data-driven decision-making with Python

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Data Visualisation in Python



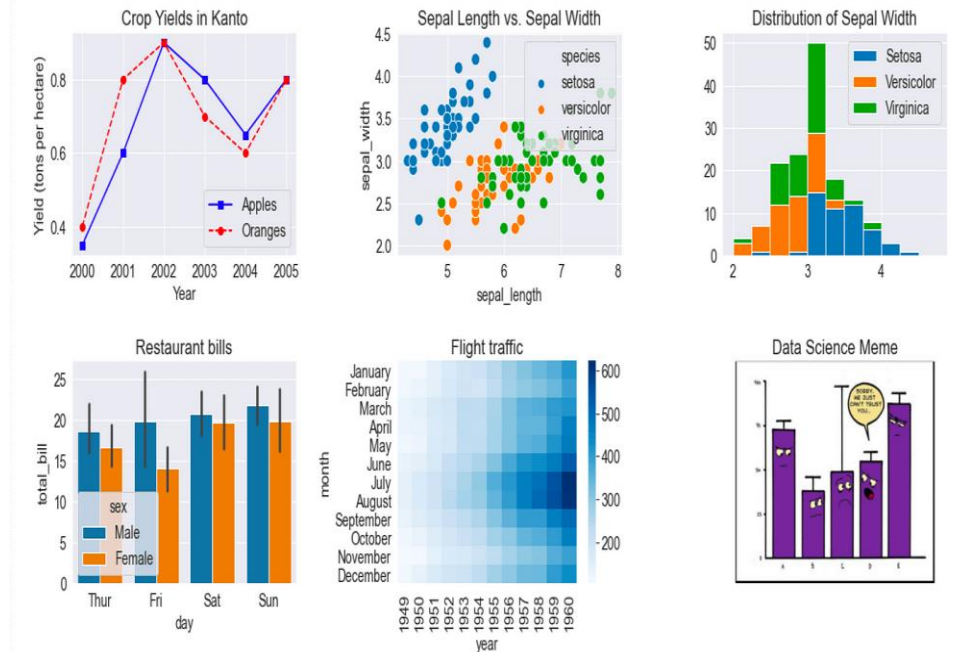
Definition: graphical representation of information using visual elements (charts, graphs, maps).



Importance: Helps quickly and clearly understand complex data.



Python Visualization Libraries: Matplotlib, Seaborn, Plotly, Pandas.



Characteristics of Good Data Visualisation

- Accurate representation
- Clear labeling and titles
- Appropriate scales and axes
- Effective color usage
- Minimal clutter
- Suitable chart type selection



Common Pitfalls in Data Visualisation

Poor labeling

Misleading
scales

Cluttered
visuals

Inappropriate
color choices

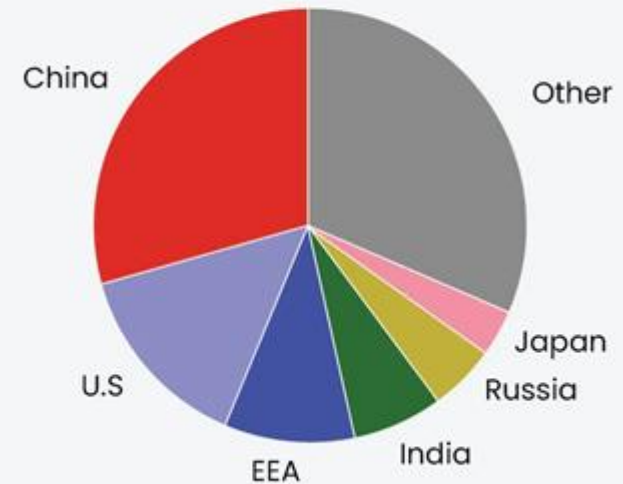
Unnecessary
effects

Wrong chart
type

Ignoring data
order

Overplotting

**Bad Data Visualization
Examples Explained**

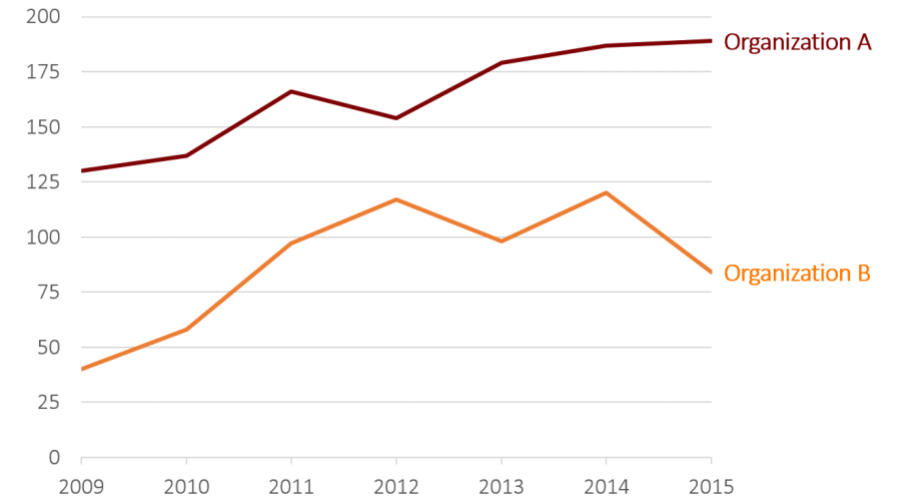


Example: Poor Labeling

Problem:
Missing
title, axes
labels, and
legend

**Python
Improvement: Use**
`plt.xlabel()`,
`plt.ylabel()`, `plt.title()`

Consequence:
Viewer confusion,
data
misinterpretation



Example: Misleading Scales

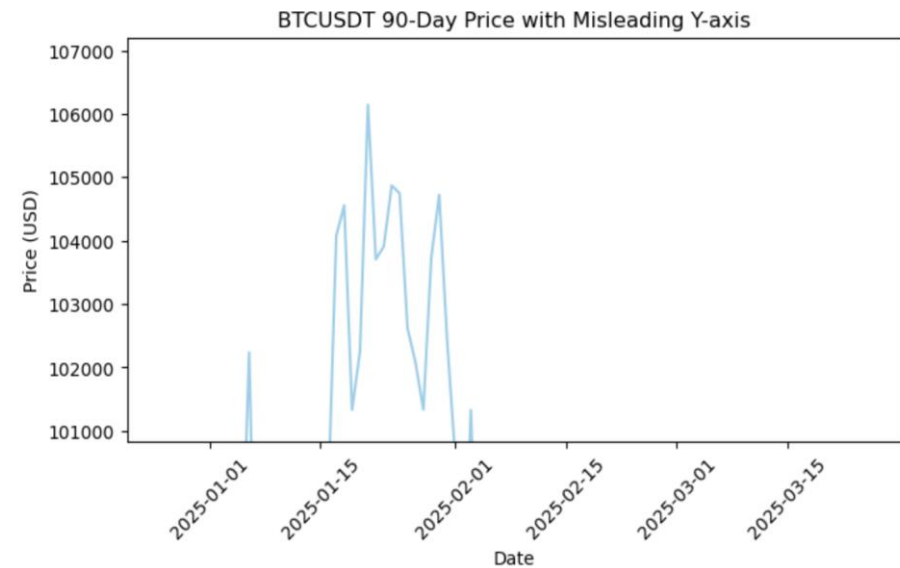
Problem: Distorting
data perception by
inappropriate
zoom



Consequence:
Exaggerated
differences



Python Solution:
Proper scaling with
`plt.xlim()`, `plt.ylim()`



Example: Cluttered Visualisations



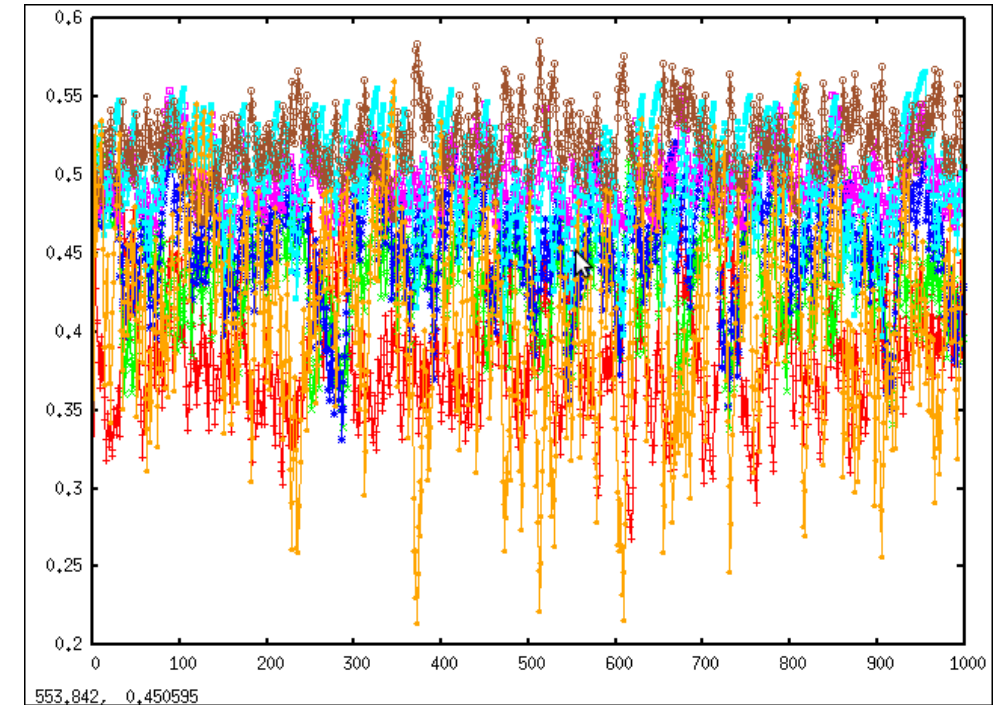
Problem: Overcrowded charts



Consequence: Difficulty identifying key data insights



Python Solution: Simplify visual elements, utilize `sns.despine()` for clarity

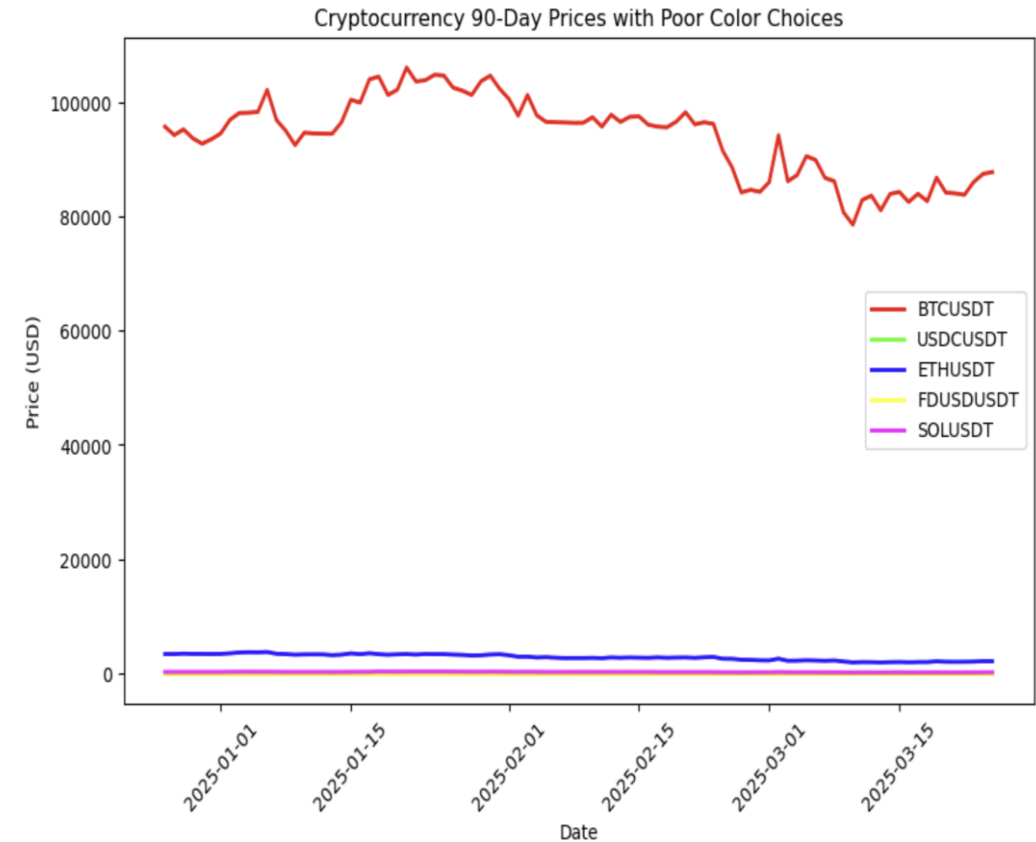


Example: Bad Color Choices

Problem: Poor contrast,
inaccessible color choices

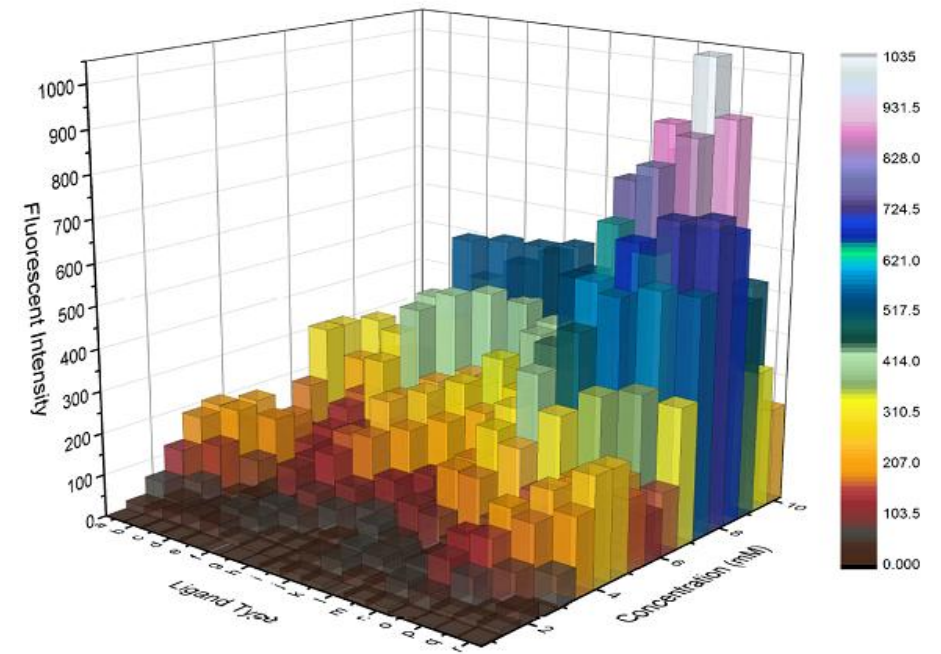
Consequence: Reduced
readability, confusion for
color-impaired viewers

Python Solution: Accessible
palettes, e.g.,
`sns.color_palette("colorblind")`



Example: Unnecessary Effects

- **Problem:** Unneeded 3D visuals
- **Consequence:** Distraction from core data insights
- **Recommendation:** Use effects sparingly and purposefully



Example: Wrong Chart Type



Problem: Misrepresentation (e.g., pie chart for time series)

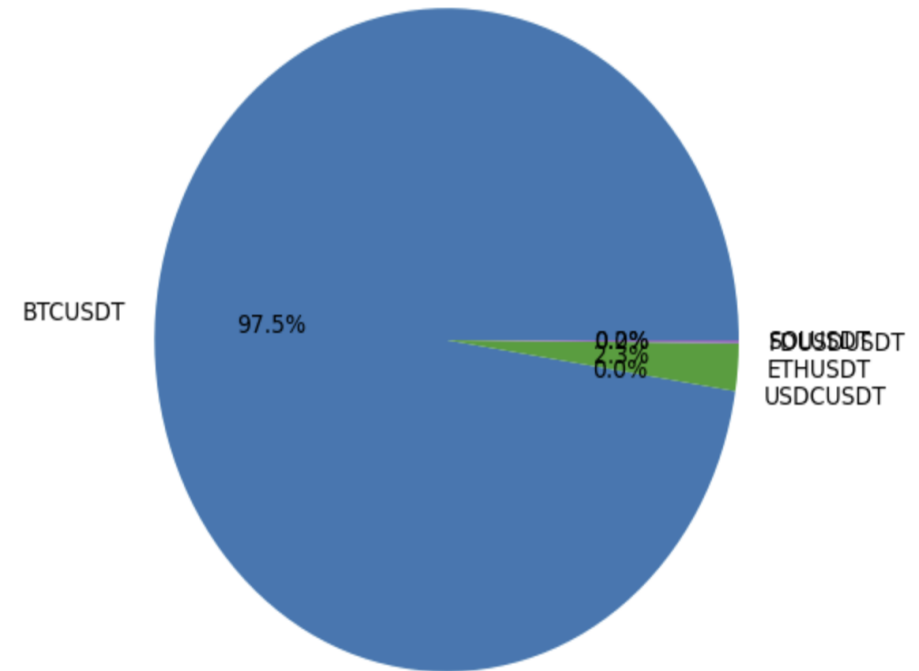


Consequence: Misinterpretation of data



Correct Approach: Select chart types matching data nature (line charts for trends)

Pie Chart of Latest Prices (Not Ideal for Time Series)



Example: Ignoring Data Order



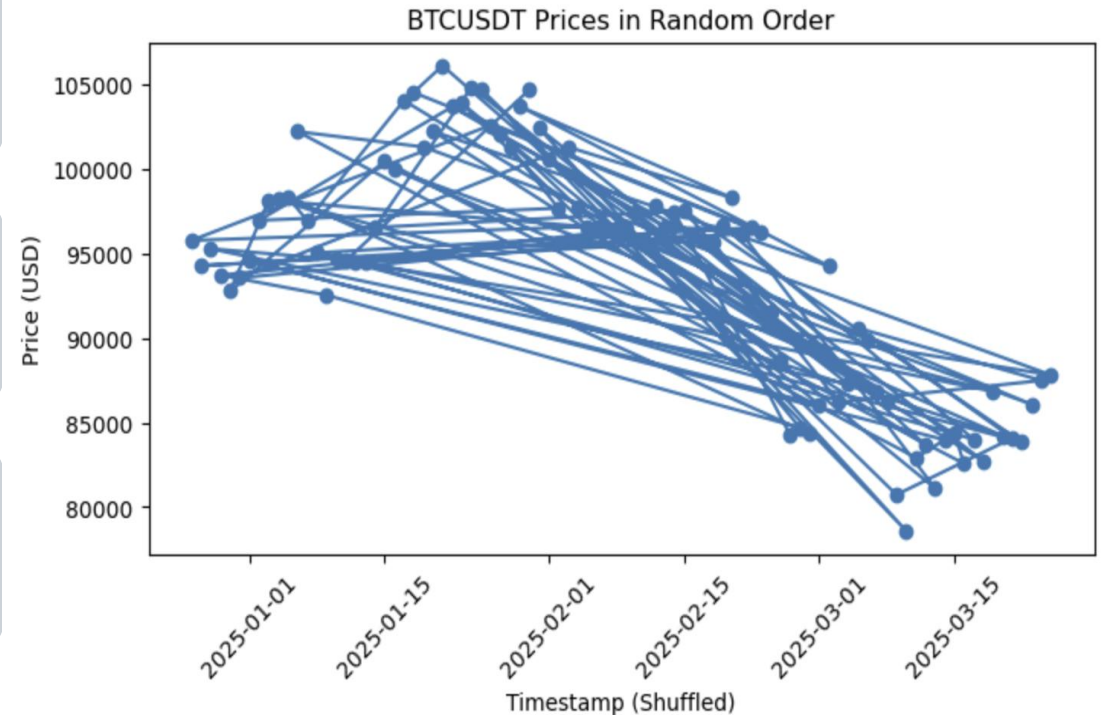
Problem: Random data order obscuring trends



Consequence: Loss of meaningful patterns



Python Solution: Ensure ordered data presentation using `df.sort_values()`



Example: Overplotting Issues

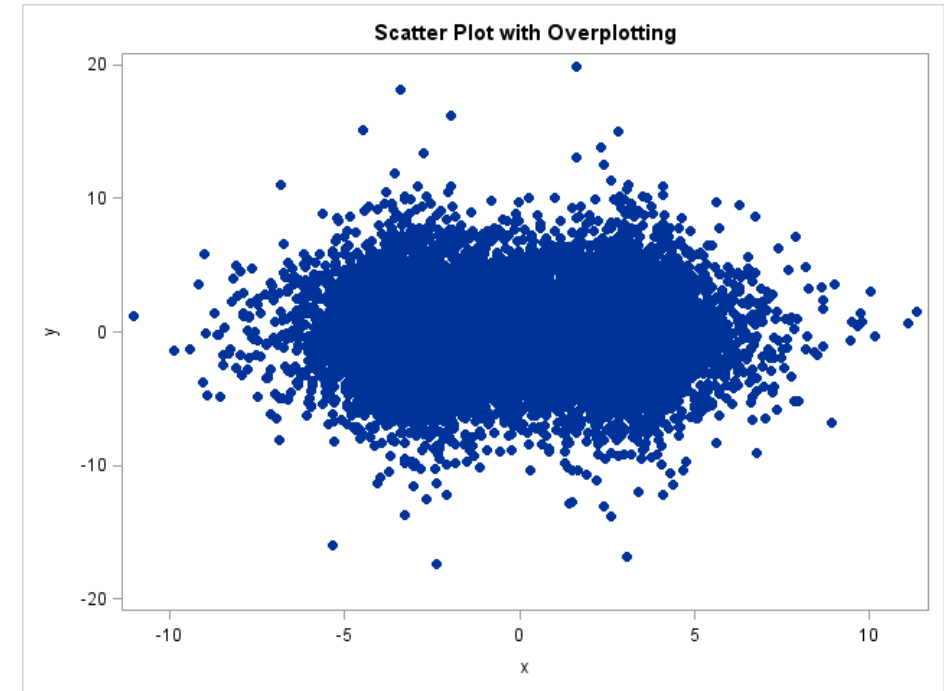
Problem: Data points overlapping excessively



Consequence: Hidden insights, unclear visualization

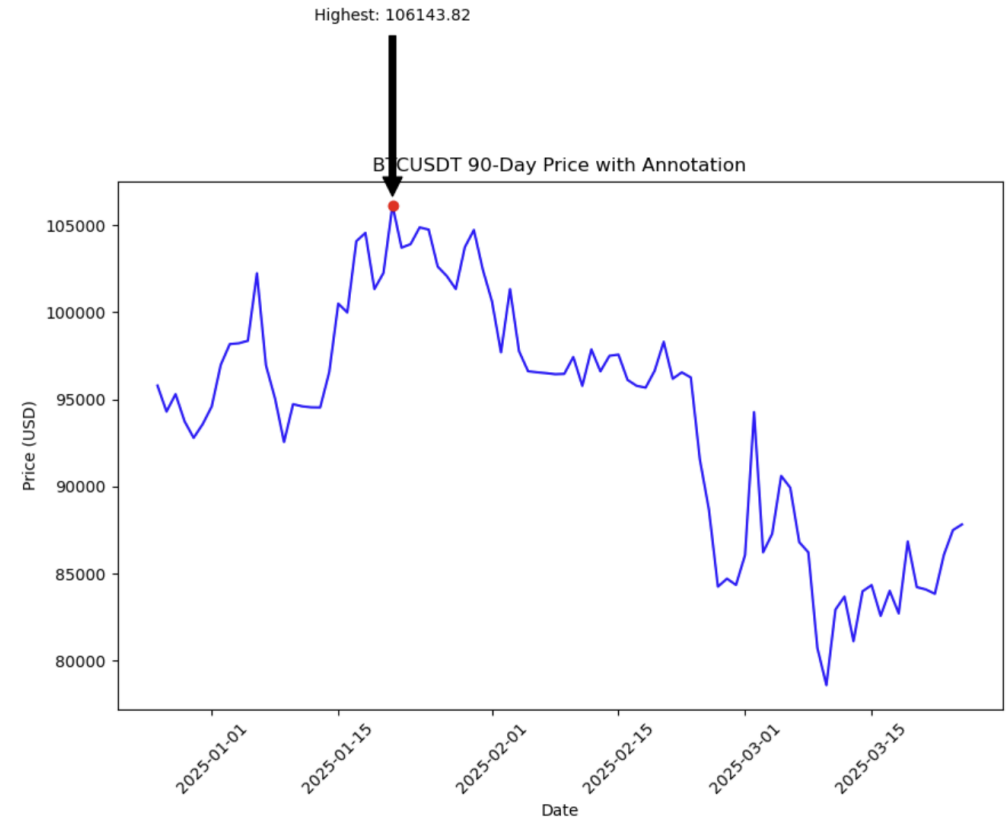


Solutions: Transparency (alpha), jittering, hexbin plots (`plt.hexbin()`), or density plots (`sns.kdeplot()`)



Enhanced Visualisation with Annotations

- **Benefit:** Highlighting key insights, improving viewer engagement
- **Example:** Annotating highest price point clearly using `plt.annotate()`



Conclusion

- Effective data visualisation aids:
 1. clear communication,
 2. supports informed decision-making,
 3. and ensures professionalism in data analytics.

