

project

December 27, 2025

1 About the Project

1.1 Overview

The project involves the statistical analysis of the official results of the Intra-School Council Elections using libraries like Pandas, Matplotlib & Seaborn and SQL.

The student council elections are held every year to choose representatives for different leadership roles in the school and the election process is managed entirely by student-led tech teams.

1.2 Elections

The elections were held in school using a custom made *Election Software* (1), a fullstack application that I designed and developed featuring:

- A frontend built with HTML, CSS, and TypeScript
- A REST API using FastAPI (Python)
- Votes securely stored in a MongoDB database

1.3 Report Generation

This report is a single Jupyter notebook exported via a custom script to HTML and styled with CSS (and then printed).

Source available at *GitHub Repo*(2).

1.4 References

1. <https://github.com/d1vij/electionsoftware>
 2. <https://github.com/d1vij/ip-proj>
-

1.5 Libraries Used

1. *Pymongo* → for querying vote documents from MongoDB server
2. *sqlite3* → for querying local SQL database

3. *pandas* → for data manipulation and analysis
4. *seaborn* → graphing
5. *matplotlib* → graphing

1.5.1 Importing Stuff

```
[1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

sns.set_theme()
```

1.5.2 Defining Constants

1.6 Structure of Data

Each vote session is stored as a single MongoDB document and the documents are segregated into collections based on the class of the student which casted the vote

A vote session represents all votes cast by a single client in one voting attempt.

Fields 1. *client* - Unique Identifier of the device that vote was casted on

2. *token*

- UUID to identify the vote session. Each *valid* vote has a corresponding unique token

3. *vote_data*

- Array of vote objects
- Each vote object contains the name of candidate and the post the vote is casted for

// example single vote document

```
{
  "_id": {
    "$oid": "68a1819ceff178ec25b66fbb"           // internal mongodb document id
  },
  "token": "b489737f-7997-430c-950f-b8c1b22f68c3", // A unique uuid4 token identifying the v
  "client": "29",                                   // Computer on which the vote was casted
  "vote_data": [                                     // Candidates voted by the voter
    {
      "name": "Abhichandra Charke",
      "post": "Captain Boy"
    },
    {
      "name": "Gauravi Zade",
      "post": "Captain Girl"
    },
    {
      "name": "Kausar Chandra",
```

```

        "post": "Vice Captain Boy"
    },
    {
        "name": "Ketaki Phalle",
        "post": "Vice Captain Girl"
    }
]
}

```

[2]: *# Fetches all the documents of all classes from the mongodb cluster*
Returns an array of dictionaries containing the feilds for name of class and
↳ array of vote sessions

```

def get_classwise_documents(
    connection_url: str, database_name: str, classes: list[str]
) -> list[dict]:
    import pymongo

    conn = pymongo.MongoClient(connection_url)
    database = conn.get_database(database_name)
    all_documents: list[dict] = []
    vote_document: dict
    for class_name in classes:
        class_documents: list[dict] = []
        collection = database.get_collection(class_name)

        for vote_document in collection.find({}):
            class_documents.append(vote_document["vote_data"]) # type: ignore

        all_documents.append({"name": class_name, "votes": class_documents})

    return all_documents

```

```

[
    {
        "name": "10A",
        "votes": [
            [
                { "name": "Aadityaraje Desai", "post": "Captain Boy" },
                { "name": "Tvisha Shah", "post": "Captain Girl" },
                { "name": "Kausar Chandra", "post": "Vice Captain Boy" },
                { "name": "Ketaki Phalle", "post": "Vice Captain Girl" }
            ],
            [
                { "name": "Abhichandra Charke", "post": "Captain Boy" },
                { "name": "Tvisha Shah", "post": "Captain Girl" },
                { "name": "Sagnik Ghosh", "post": "Vice Captain Boy" },
                { "name": "Ketaki Phalle", "post": "Vice Captain Girl" }
            ]
        ]
    }
]

```

```
[
  { "name": "Aadityaraje Desai", "post": "Captain Boy" },
  { "name": "Tvisha Shah", "post": "Captain Girl" },
  { "name": "Avaneesh Mahalle", "post": "Vice Captain Boy" },
  { "name": "Trisha Kandpal", "post": "Vice Captain Girl" }
],
[
  { "name": "Aadityaraje Desai", "post": "Captain Boy" },
  { "name": "Gauravi Zade", "post": "Captain Girl" },
  { "name": "Kausar Chandra", "post": "Vice Captain Boy" },
  { "name": "Riya Shirole", "post": "Vice Captain Girl" }
]
...
```

```
[3]: CONNECTION_URL = "mongodb+srv://vermadivij:elections@cluster1.#####.mongodb.
↳net/"
CLASSES = ['10A', '10B', '10C', '10D', '10E', '10F', '10G', '10H', '10I',
↳'10J', '11A', '11B', '11C', '11D', '11E', '12A', '12B', '12C', '12D', '9A',
↳'9B', '9C', '9D', '9E', '9F', '9G', '9H', '9I', '9J', 'absentees',
↳'candidates'] # fmt: off
CLUSTER="votes"

def download_results():
    import json
    documents = get_classwise_documents(CONNECTION_URL, "votes", CLASSES)
    with open("votes.json", "w+") as file:
        file.write(json.dumps(documents))
```

1.7 Compiling Data

```
[4]: # fmt: off
candidate_data = {
    "Captain Boy": [ "Aadityaraje Desai", "Abhichandra Charke", "Praneel
↳Deshmukh", "Rachit Srivastava", ],
    "Captain Girl": [ "Tvisha Shah", "Gauravi Zade", "Kirthika Jayachander",
↳"Naisha Rastogi", ],
    "Vice Captain Boy": [ "Kausar Chandra", "Sagnik Ghosh", "Avaneesh Mahalle",
↳"Krishna Yadav", "Viren Jadhav", ],
    "Vice Captain Girl": [ "Ketaki Phalle", "Trisha Kandpal", "Riya Shirole",
↳"Kavya Mehta", "Sumedha Vaidya", ],
}
# fmt: on
import json
import pandas as pd

def get_classwise_documents_from_local(votes_json: str) -> list[dict]:
    with open(votes_json) as file:
        return json.loads(file.read())
```

```

def calculate_votes(votes_json: str) -> pd.DataFrame:
    classwise_votes = get_classwise_documents_from_local(votes_json)
    votes_df = pd.DataFrame(
        [
            {
                "class": _class["name"],
                "candidate_name": candidate["name"],
                "post": candidate["post"],
            }
            for _class in classwise_votes
            for votes in _class["votes"]
            for candidate in votes
        ],
        columns=["class", "candidate_name", "post"], # type: ignore
    )
    return votes_df

votes_df = calculate_votes("votes.json")
print(votes_df.head(10))

```

	class	candidate_name	post
0	10A	Aadityaraje Desai	Captain Boy
1	10A	Tvisha Shah	Captain Girl
2	10A	Kausar Chandra	Vice Captain Boy
3	10A	Ketaki Phalle	Vice Captain Girl
4	10A	Abhichandra Charke	Captain Boy
5	10A	Tvisha Shah	Captain Girl
6	10A	Sagnik Ghosh	Vice Captain Boy
7	10A	Ketaki Phalle	Vice Captain Girl
8	10A	Aadityaraje Desai	Captain Boy
9	10A	Tvisha Shah	Captain Girl

```

[5]: # Classwise dataframes
classwise_grouped = votes_df.groupby("post")

cb = votes_df.loc[classwise_grouped.groups["Captain Boy"]].drop("post", axis=1).
    ↪reset_index(drop=True)
cg = votes_df.loc[classwise_grouped.groups["Captain Girl"]].drop("post",
    ↪axis=1).reset_index(drop=True)
vcb = votes_df.loc[classwise_grouped.groups["Vice Captain Boy"]].drop("post",
    ↪axis=1).reset_index(drop=True)
vcg = votes_df.loc[classwise_grouped.groups["Vice Captain Girl"]].drop("post",
    ↪axis=1).reset_index(drop=True)

postwise_votes_df = {

```

```

    "captain_boy": cb,
    "captain_girl": cg,
    "vice_captain_boy": vcb,
    "vice_captain_girl": vcg,
}

```

1.8 SQL Querying

SqliteDatabase is a wrapper around the `sqlite3` library which allows for quick and hasslefree SQL querying of the data.

```

[6]: import sqlite3
from sqlite3 import Connection, Cursor
from typing import Any, Literal

class SqliteDatabase:
    def __init__(self, database: str):
        self.database = database
        self.conn: Connection | None = None
        self.cursor: Cursor | None = None

    def __enter__(self):
        try:
            self.conn = sqlite3.connect(self.database)
            self.cursor = self.conn.cursor()
        except Exception as e:
            print(f"Error occured in connecting to the database {self.database}."
↪ Error Details: {e}")

        return self.query

    def __exit__(self, exc_type, exc, tb):
        assert self.conn is not None
        assert self.cursor is not None

        self.cursor.close()
        self.conn.close()

        return False # dont suppress the error

    def query(
        self,
        query: str,
        *,
        is_updation=False, # is the current query contains some kind of ↵
↪ updation ?? Doesnt return anything if true

```

```

return_rows: None | Literal["str"] | Literal["tuple"] = None,
table_heading: str | None = None, # Title printed before printing
↪output
) -> None | tuple[tuple[str, ...], ...]:
    assert self.conn is not None
    assert self.cursor is not None

    try:
        results = self.cursor.execute(query)
        self.conn.commit()
    except Exception as err:
        print("** Row / Column names with spaces should be enclosed within
↪quotes **")
        raise err

    if is_updation:
        return

    rows: list[Any] = results.fetchall()
    columns_headers: tuple[str, ...] = tuple(str(col[0]) for col in results.
↪description)

    lines: tuple[tuple[str, ...], ...] = tuple((columns_headers, *rows))

    if return_rows == "tuple":
        return lines
    elif return_rows is None:
        # printing table header if provided
        if table_heading is not None:
            print(table_heading)

        # Finding max column width
        column_widths: list[int] = []

        for col_idx in range(len(lines[0])):
            widths = []
            for row_idx in range(len(lines)):
                widths.append(len(str(lines[row_idx][col_idx])))
            column_widths.append(max(widths))

        # Printing column headers
        border_top_bottom = "+" + "-" * (sum(column_widths) + 3 *
↪len(column_widths) - 1) + "+"
        print(border_top_bottom)
        print("| ", end="")
        for idx, col_label in enumerate(lines[0]):
            print(str(col_label).ljust(column_widths[idx]), end=" | ")

```

```

print()
print(border_top_bottom)

for row in rows:
    print("| ", end="")
    for idx, col_val in enumerate(row):
        print(str(col_val).ljust(column_widths[idx]), end=" | ")
    print()

print(border_top_bottom)
return None

```

1.9 Creating post dataframes and saving them to SQLite database

```

[7]: with SQLiteDatabase("votes.db") as query:
    for post_name, post_df in postwise_votes_df.items():
        query(f"drop table if exists {post_name};", is_updation=True)
        query(
            f"create table {post_name} (class varchar(255), candidate_name_␣
↪varchar(255));",
            is_updation=True,
        )

        rows = []
        for (idx, row) in post_df.iterrows():
            (class_name, candidate_name) = (row["class"], row["candidate_name"])
            rows.append(f"('{class_name}', '{candidate_name}')"

        query(f"insert into {post_name} values" + ",".join(rows),␣
↪is_updation=True)
        query(f"select * from {post_name} limit 10", table_heading=post_name)

```

captain_boy

```

+-----+
| class | candidate_name |
+-----+
| 10A   | Aadityaraje Desai |
| 10A   | Abhichandra Charke |
| 10A   | Aadityaraje Desai |
| 10A   | Aadityaraje Desai |
| 10A   | Abhichandra Charke |
| 10A   | Abhichandra Charke |
| 10A   | Praneel Deshmukh |
| 10A   | Aadityaraje Desai |
| 10A   | Praneel Deshmukh |
| 10A   | Aadityaraje Desai |

```



```

+-----+
captain_girl
+-----+
| class | candidate_name |
+-----+
| 10A   | Tvisha Shah    |
| 10A   | Tvisha Shah    |
| 10A   | Tvisha Shah    |
| 10A   | Gauravi Zade   |
| 10A   | Gauravi Zade   |
| 10A   | Gauravi Zade   |
| 10A   | Gauravi Zade   |
| 10A   | Kirthika Jayachander |
| 10A   | Gauravi Zade   |
| 10A   | Naisha Rastogi |
+-----+
vice_captain_boy
+-----+
| class | candidate_name |
+-----+
| 10A   | Kausar Chandra |
| 10A   | Sagnik Ghosh   |
| 10A   | Avaneesh Mahalle |
| 10A   | Kausar Chandra |
| 10A   | Krishna Yadav  |
| 10A   | Kausar Chandra |
| 10A   | Krishna Yadav  |
| 10A   | Avaneesh Mahalle |
| 10A   | Avaneesh Mahalle |
| 10A   | Krishna Yadav  |
+-----+
vice_captain_girl
+-----+
| class | candidate_name |
+-----+
| 10A   | Ketaki Phalle  |
| 10A   | Ketaki Phalle  |
| 10A   | Trisha Kandpal |
| 10A   | Riya Shirode   |
| 10A   | Ketaki Phalle  |
| 10A   | Kavya Mehta    |
| 10A   | Ketaki Phalle  |
| 10A   | Trisha Kandpal |
| 10A   | Trisha Kandpal |
| 10A   | Trisha Kandpal |
+-----+

```

1.10 Statistical Analysis

1.10.1 Total Votes across all classes

```
[8]: fig, axes = plt.subplots(2,2, figsize=(15,10))
positions = [(0,0), (0,1), (1,0), (1,1)]

for idx, (post_name, post_df) in enumerate(postwise_votes_df.items()):
    with SqliteDatabase("votes.db") as query:
        query(
            f"""
            select candidate_name as Name, count(*) as Votes
            from {post_name}
            group by candidate_name
            order by Votes desc;
            """,
            table_heading="Total Votes for - " + post_name
        )
    print()
    votes = post_df.groupby("candidate_name").size()
    sns.barplot(votes, ax=axes[positions[idx]]) #type: ignore
    axes[positions[idx]].set_title(post_name)
plt.tight_layout()
plt.show()
print()
```

Total Votes for - captain_boy

+-----+		
Name	Votes	
+-----+		
Praneel Deshmukh	151	
Aadityaraje Desai	148	
Abhichandra Charke	70	
Rachit Srivastava	16	
+-----+		

Total Votes for - captain_girl

+-----+		
Name	Votes	
+-----+		
Gauravi Zade	252	
Naisha Rastogi	72	
Tvisha Shah	41	
Kirthika Jayachander	20	
+-----+		

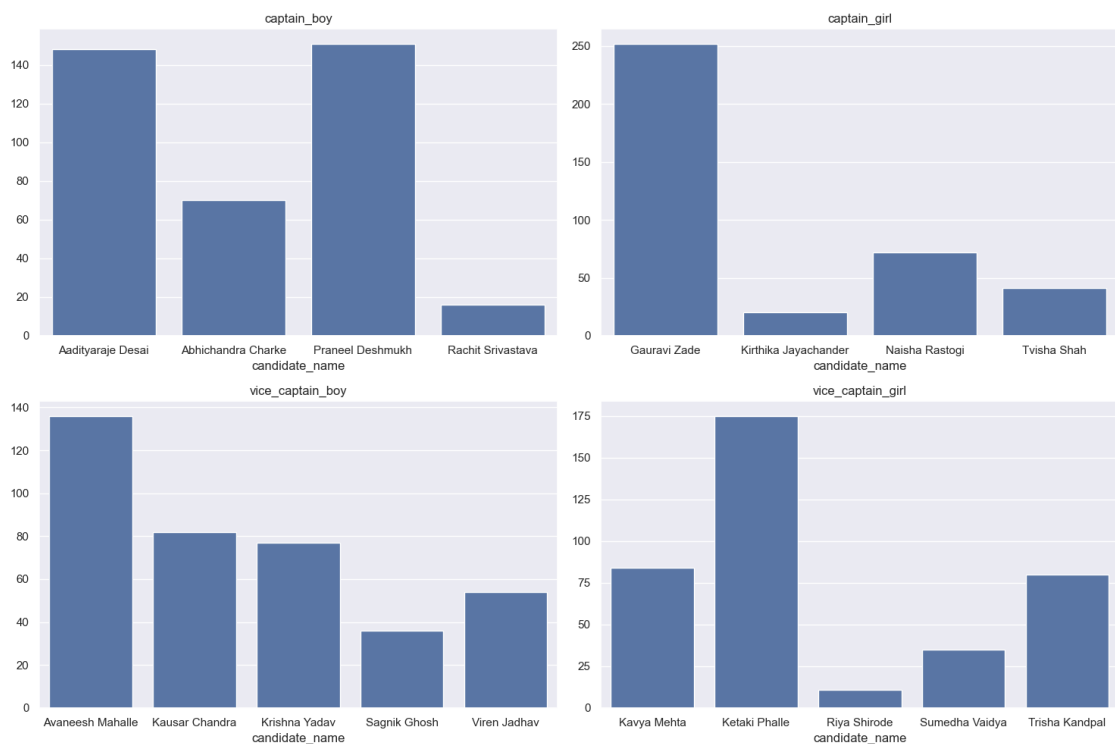
Total Votes for - vice_captain_boy

+-----+		
Name	Votes	

+-----+		
Avaneesh Mahalle	136	
Kausar Chandra	82	
Krishna Yadav	77	
Viren Jadhav	54	
Sagnik Ghosh	36	
+-----+		

Total Votes for - vice_captain_girl

+-----+		
Name	Votes	
+-----+		
Ketaki Phalle	175	
Kavya Mehta	84	
Trisha Kandpal	80	
Sumedha Vaidya	35	
Riya Shirode	11	
+-----+		



1.11 Candidate Popularity trends

```
[9]: # candidate popularity trends - comparing candidate performances across classes

from matplotlib.ticker import MultipleLocator

def plot_popularity_trends(post_name: str, post_df: pd.DataFrame):

    # extracting rows belonging to a particular class from the post's dataframe
    ↪ using regular expressions
    class_wise_dataframes = [
        post_df.loc[post_df["class"].str.contains(_re)]
        for _re in [r"9\w", r"10\w", r"11\w", r"12\w"] # <--- regex btw
    ]

    # dividing the plot into 4 subplots
    fig, axes = plt.subplots(2, 2, figsize=(15, 7))

    subplot_positions = [
        (0, 0),
        (0, 1),
        (1, 0),
        (1, 1),
    ] # since there are only 4 classes / subplots
    linestyle = [":", "-", "--", "-.", "solid"]

    for idx in range(4):
        pos = subplot_positions[idx]

        class_df = class_wise_dataframes[idx]

        # series of all sections
        sections = class_df["class"].unique()
        candidates = class_df["candidate_name"].unique()

        # Group all the votes by class, then candidate
        # Create a new column with the total size of each group (ie votes in
        ↪ that class) and name it votes
        classwise_votes = class_df.groupby(["class", "candidate_name"]).size().
        ↪ reset_index(name="votes")

        # ^^^ Groupby forms a series with a
        ↪ multi-index
        for idx, candidate_name in enumerate(candidates):
            candidate_votes = (
                classwise_votes
                # Select all the votes for a given candidate
```

```

        .loc[classwise_votes["candidate_name"] == candidate_name]
        # Set the "class" column as the index and take only the
↪ votes column

        .set_index("class")["votes"]
        # Ensure that the candidate has entries for each section,
        # fill the missing ones with 0
        .reindex(sections, fill_value=0)
    )
    # plotting a subplot for each class
    axes[pos].plot(
        sections,
        candidate_votes,
        label=candidate_name,
        linestyle=linestyles[idx],
    )

    axes[pos].set_xlabel("class")
    axes[pos].set_ylabel("Votes")

    # axes[pos].set_ylim(0, post_df.max().max() + 1)

    # values on y-axis would have a difference of 2
    axes[pos].yaxis.set_major_locator(MultipleLocator(2))

fig.suptitle(post_name, fontsize=32)

# setting a common legend for the whole plot
handles, labels = axes[0, 0].get_legend_handles_labels()
fig.legend(handles, labels, loc="upper right", ncols=2, fontsize=15)

plt.show()

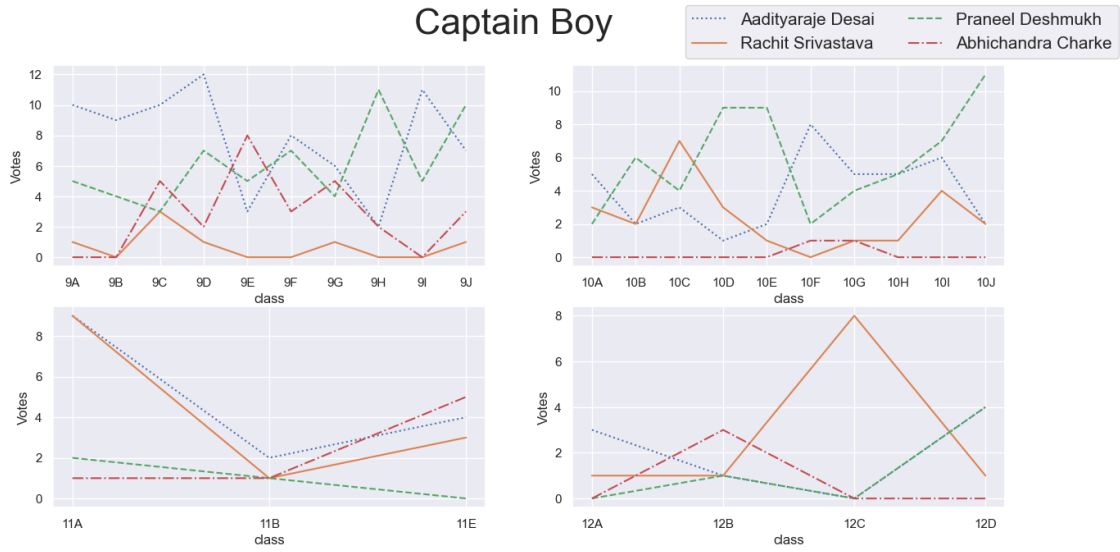
```

```

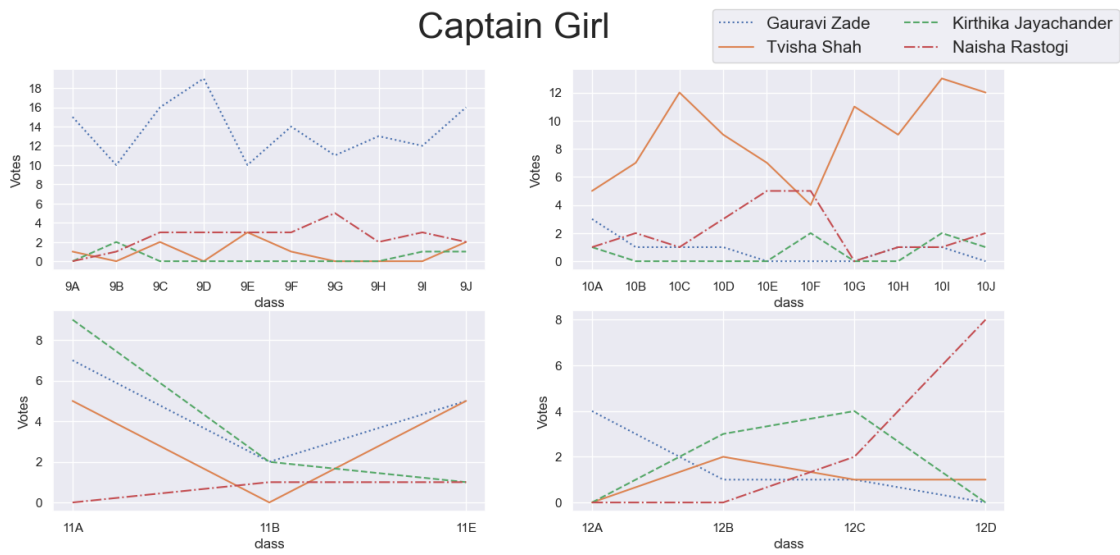
plot_popularity_trends("Captain Boy", cb)
plot_popularity_trends("Captain Girl", cg)
plot_popularity_trends("Vice Captain Boy", vcb)
plot_popularity_trends("Vice Captain Girl", vcg)

```

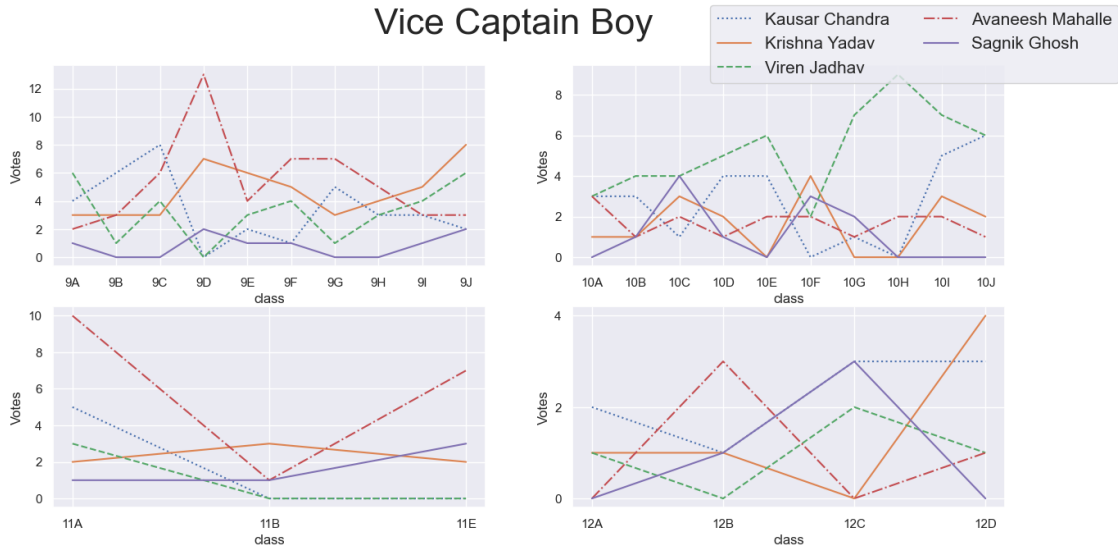
Captain Boy



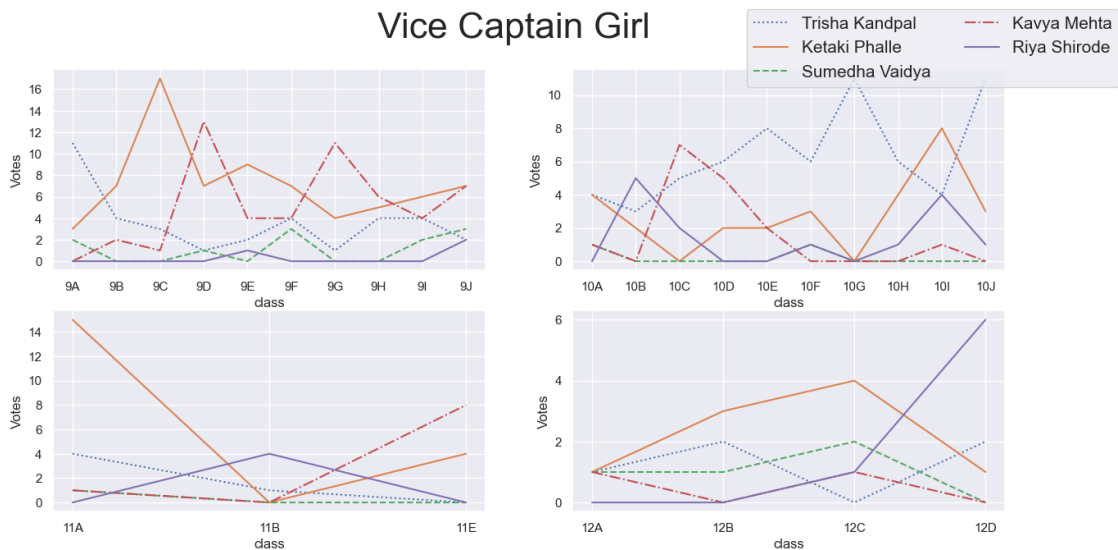
Captain Girl



Vice Captain Boy



Vice Captain Girl



1.12 Majority Share

The share in percent of classes in which a candidate has majority

```
[10]: total_classes = len(cb["class"].unique())

fig, axes = plt.subplots(2, 2, figsize=(10, 7), constrained_layout=True)
fig.suptitle("Share (percent) of Classes in which a Candidate has a majority ↵",
             ↵", fontsize=20)
```

```

subplot_positions = [
    (0, 0),
    (0, 1),
    (1, 0),
    (1, 1),
]

colors = plt.cm.copper_r(np.linspace(0,0.50,5)) # type: ignore

for idx, (post_name, post_df) in enumerate(postwise_votes_df.items()):
    pos = subplot_positions[idx]
    # kitne classes me each candidate has the maximum votes

    votes = post_df.value_counts(["class", "candidate_name"]).
    ↪reset_index(name="votes")
    winners = votes.loc[votes.groupby("class")["votes"].idxmax()]
    count_series = winners.groupby("candidate_name").size()
    percents_series = count_series / total_classes

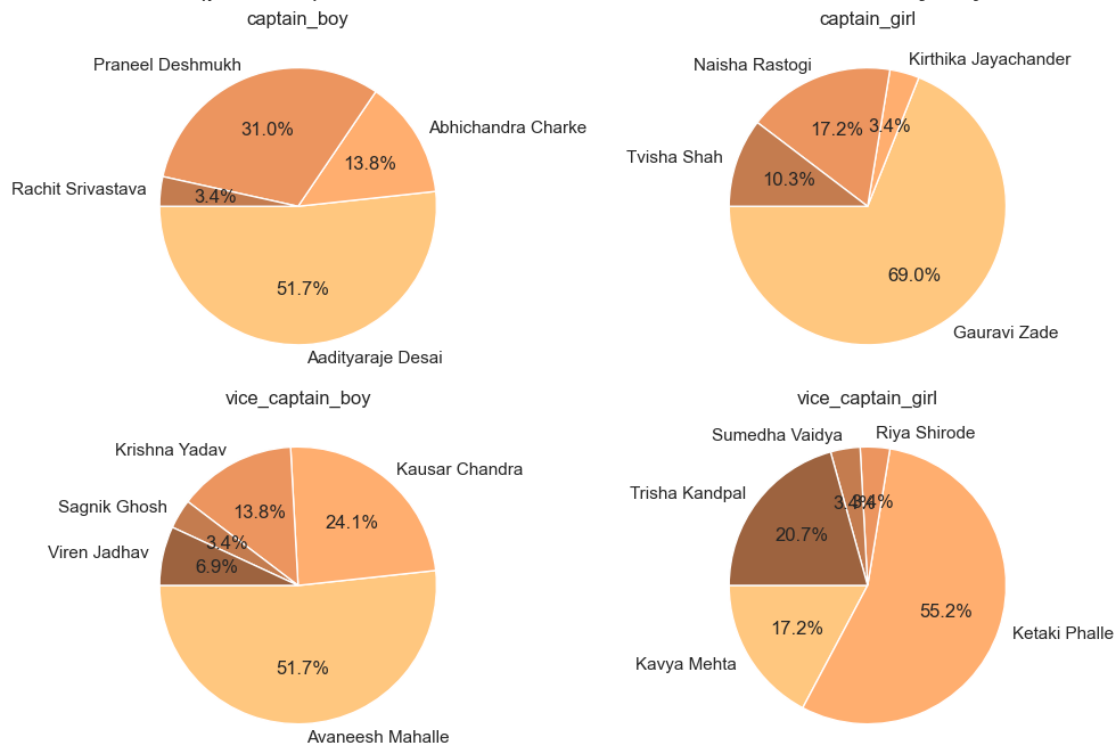
    max_val = percents_series.max()

    axes[pos].set_title(post_name)
    axes[pos].pie(
        percents_series,
        labels=percents_series.index.map(
            lambda name: name.replace("_", " ")
        ),
        autopct="%1.1f%%",
        startangle=180,
        colors=colors,
    )
    axes[pos].set(aspect='equal')

plt.show()

```


Share (percent) of Classes in which a Candidate has a majority



Analyzes whether voters who supported one candidate also tended to support another.

Working:

1. Builds a `vote session` dataframe recording which candidates were chosen in each voting session.

Example

	Captain Boy	Captain Girl	Vice Captain Boy	Vice Captain Girl	class
25	Praneel Deshmukh	Gauravi Zade	Viren Jadhav	Kavya Mehta	10C
26	Aadityaraje Desai	Gauravi Zade	Avaneesh Mahalle	Kavya Mehta	10C
27	Abhichandra Charke	Gauravi Zade	Avaneesh Mahalle	Ketaki Phalle	10C
28	Praneel Deshmukh	Gauravi Zade	Sagnik Ghosh	Sumedha Vaidya	10C
29	Abhichandra Charke	Gauravi Zade	Sagnik Ghosh	Sumedha Vaidya	10C
30	Abhichandra Charke	Gauravi Zade	Avaneesh Mahalle	Ketaki Phalle	10C

2. Constructs a co-occurrence matrix showing how often Candidate B was voted when Candidate A was voted.

Example

	Gauravi Zade	Kirthika Jayachander	Naisha Rastogi
Avaneesh Mahalle	94	4	26

Krishna Yadav	54	3	14
Viren Jadhav	33	2	11
Ketaki Phalle	116	5	38
Trisha Kandpal	52	5	15
Riya Shirode	5	3	0
Kavya Mehta	52	3	18
Sumedha Vaidya	27	4	1

3. Normalizes it into a probability matrix to estimate the likelihood of co-support between candidates. Each row of the co-occurrence matrix is divided by the total votes in that row. This converts raw counts into conditional probabilities, i.e., the chance of Candidate B being voted given that Candidate A was voted.
4. Visualizes both matrices using heatmaps — one for raw counts, the other for probabilities.

```
[11]: from itertools import combinations

all_candidates = [
    candidate
    for post_candidates in candidate_data.values()
    for candidate in post_candidates
]

classwise_votes = get_classwise_documents_from_local("votes.json")

# Which candidates were voted together in a particular session
vote_sessions = pd.DataFrame(columns=candidate_data.keys())

for _class in classwise_votes:
    for session in _class["votes"]:
        voted = {}
        for vote in session:
            voted[vote["post"]] = vote["name"]
        vote_sessions.loc[len(vote_sessions)] = voted

print(vote_sessions.head())

# co-occurrence matrix is the matrix showing how many times candidate B was
↳ voted when candidate A was voted
# co-occurrence matrix would be N * N where N are the total number of candidates
↳ across all posts (18 * 18 for this case)
co_occurrence_matrix = pd.DataFrame(0, index=all_candidates,
    ↳ columns=all_candidates)

# updating co-occurrence matrix
for (_, session) in vote_sessions.iterrows():
    for (candidate_A, candidate_B) in combinations(session.to_list(), 2):
        co_occurrence_matrix.at[candidate_A, candidate_B] += 1
```

```

co_occurance_matrix.at[candidate_B, candidate_A] += 1

# creating conditional probability matrix
# conditional probability matrix is created by normalizing columns of
↳ co-occurance matrix
# normalizing means dividing each row of co-occurance matrix by the total votes
↳ in that row
# the matrix gives the probability of person b (x axis) being voted when person
↳ A (y axis) was voted
probability_matrix = co_occurance_matrix.div(co_occurance_matrix.sum(axis=1),
↳ axis=0)

# ---- first plot ----
fig1, ax1 = plt.subplots(figsize=(12, 8))
ax1.set_title(
    "Co-occurance plot - Number of times person A got voted when person B was
↳ voted",
    size=16,
)
sns.heatmap(
    co_occurance_matrix,
    cmap="viridis",
    vmin=0,
    annot=True,
    ax=ax1,
    fmt=".0f",
    cbar_kws={"label": "Count"},
)
ax1.set_ylabel("Person A", size=12)
ax1.set_xlabel("Person B", size=12)
ax1.set_xticklabels(ax1.get_xticklabels(), rotation=90)
ax1.set_yticklabels(ax1.get_yticklabels(), rotation=0)
plt.show()

# ---- second plot ----
fig2, ax2 = plt.subplots(figsize=(12, 8))
ax2.set_title(
    "Probability plot - Probability (in percent) of person B getting voted when
↳ person A was voted",
    size=16,
)
sns.heatmap(
    probability_matrix * 100,
    cmap="viridis",
    vmin=0,
    annot=True,

```

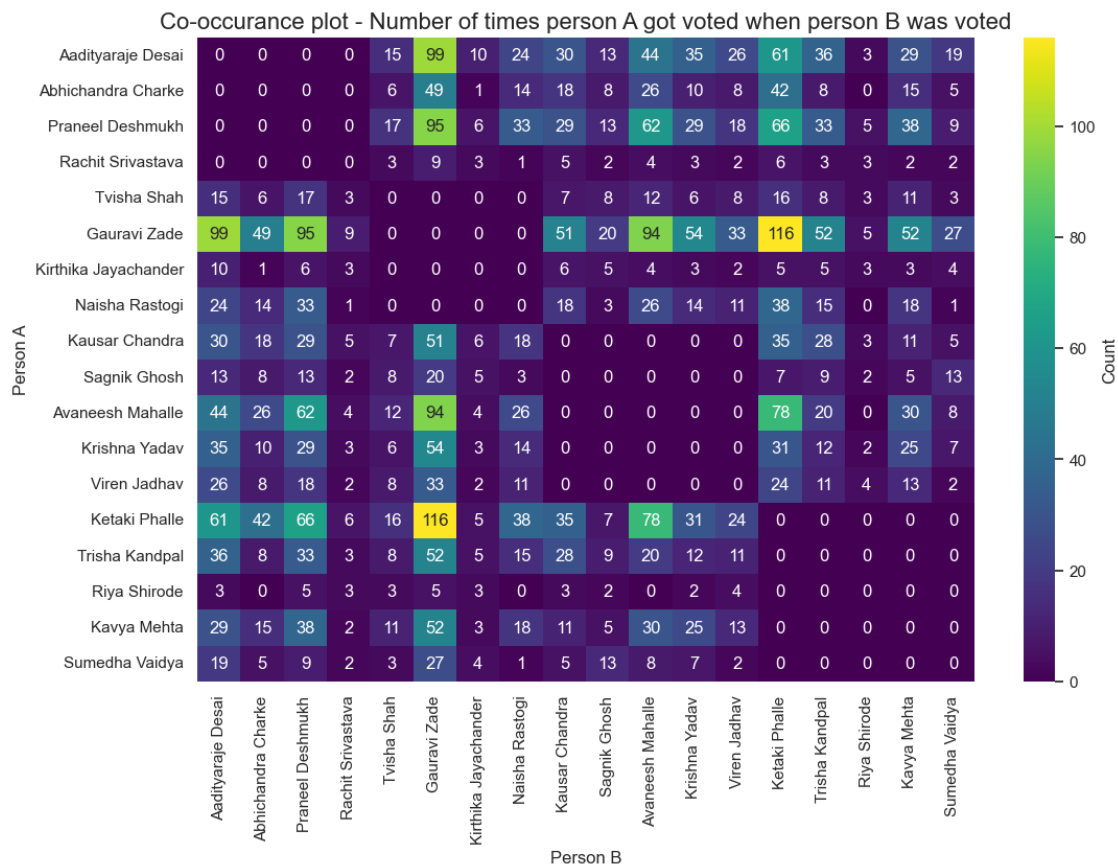
```

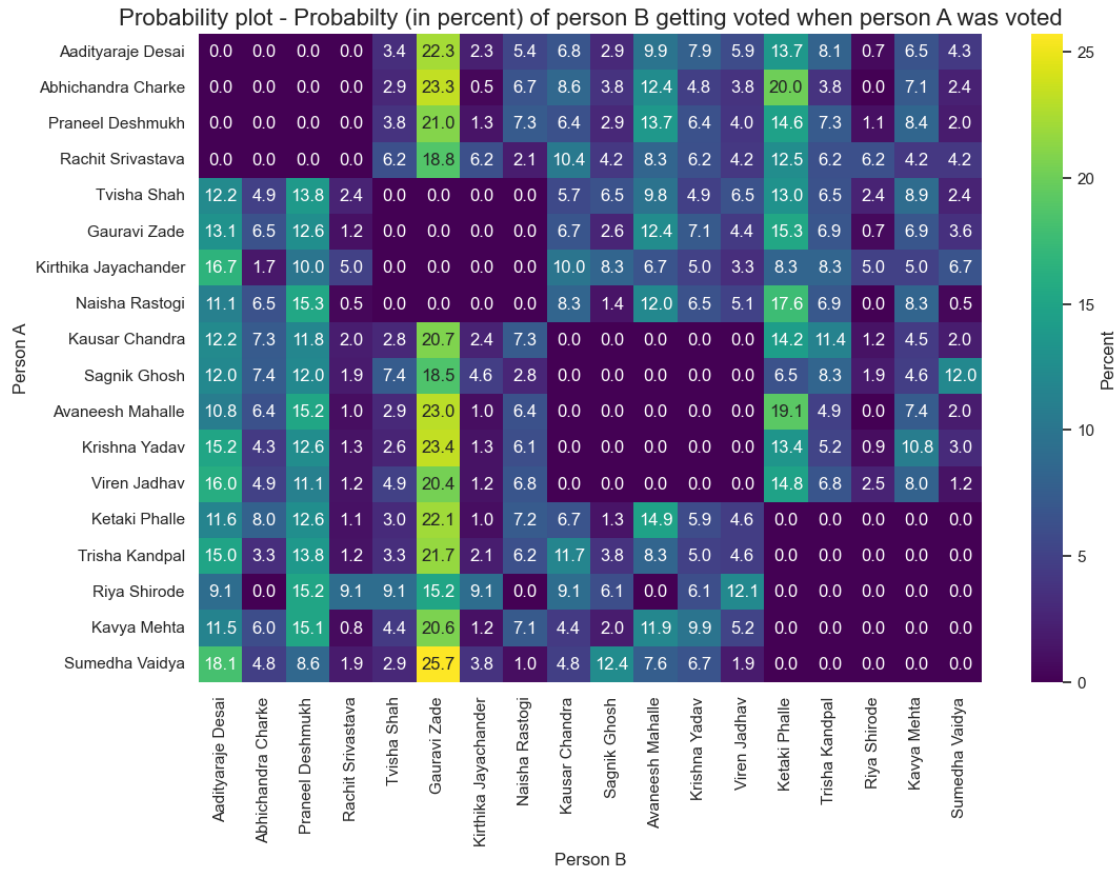
ax=ax2,
fmt=".1f",
cbar_kws={"label": "Percent"},
)
ax2.set_ylabel("Person A", size=12)
ax2.set_xlabel("Person B", size=12)
ax2.set_xticklabels(ax2.get_xticklabels(), rotation=90)
ax2.set_yticklabels(ax2.get_yticklabels(), rotation=0)
plt.show()

# the percents here for a column dont add up to 100 cuz they are mutually
↳ exclusive events

```

	Captain Boy	Captain Girl	Vice Captain Boy	Vice Captain Girl
0 Aadityaraje Desai	Tvisha Shah	Kausar Chandra	Ketaki Phalle	
1 Abhichandra Charke	Tvisha Shah	Sagnik Ghosh	Ketaki Phalle	
2 Aadityaraje Desai	Tvisha Shah	Avaneesh Mahalle	Trisha Kandpal	
3 Aadityaraje Desai	Gauravi Zade	Kausar Chandra	Riya Shirole	
4 Abhichandra Charke	Gauravi Zade	Krishna Yadav	Ketaki Phalle	





1.13 Strongest & Weakest Allies

Which candidates appeared most often (or least often) alongside another candidate during the elections.

For each candidate, the candidate with most and least co-occurrence probability is the strongest and weakest ally respectively.

```
[12]: # Use row indexes for all comparisons

strongest_ally_series = pd.Series(name="Strongest Ally")
weakest_ally_series = pd.Series(name="Weakest Ally")

for post_name, same_post_candidates in candidate_data.items():
    for name in same_post_candidates:
        # extracting the row which gives co-occurrence probability for a candidate
        cps = probability_matrix.loc[name]

        # removing values of all the candidates in the same post
```

```

candidate_probability_series = cps[~cps.index.
↳isin(same_post_candidates)]

strongest_ally_series[name] = candidate_probability_series.idxmax()
weakest_ally_series[name] = candidate_probability_series.idxmin()

# concat based on similar rows
summary = pd.concat([strongest_ally_series, weakest_ally_series], axis=1)
print(summary.sort_values(by=list(summary.columns)))

# strongest ally is the candidate who is most likely to be voted when a
↳candidate is voted
# weakest ally is the candidate who is least likely to be voted when a
↳candidate is voted

```

	Strongest Ally	Weakest Ally
Kirthika Jayachander	Aadityaraje Desai	Abhichandra Charke
Ketaki Phalle	Gauravi Zade	Kirthika Jayachander
Rachit Srivastava	Gauravi Zade	Naisha Rastogi
Sumedha Vaidya	Gauravi Zade	Naisha Rastogi
Sagnik Ghosh	Gauravi Zade	Rachit Srivastava
Viren Jadhav	Gauravi Zade	Rachit Srivastava
Trisha Kandpal	Gauravi Zade	Rachit Srivastava
Kavya Mehta	Gauravi Zade	Rachit Srivastava
Aadityaraje Desai	Gauravi Zade	Riya Shirole
Abhichandra Charke	Gauravi Zade	Riya Shirole
Praneel Deshmukh	Gauravi Zade	Riya Shirole
Kausar Chandra	Gauravi Zade	Riya Shirole
Avaneesh Mahalle	Gauravi Zade	Riya Shirole
Krishna Yadav	Gauravi Zade	Riya Shirole
Gauravi Zade	Ketaki Phalle	Riya Shirole
Naisha Rastogi	Ketaki Phalle	Riya Shirole
Riya Shirole	Praneel Deshmukh	Abhichandra Charke
Tvisha Shah	Praneel Deshmukh	Rachit Srivastava

```

[13]: # mean co-support - mean of all conditional probabilities across all candidates
probability_matrix.mean() * 100

```

```

[13]: Aadityaraje Desai      10.255480
      Abhichandra Charke    3.995533
      Praneel Deshmukh     9.970875
      Rachit Srivastava    1.704038
      Tvisha Shah          3.314564
      Gauravi Zade         16.480611
      Kirthika Jayachander  2.111771
      Naisha Rastogi       4.018980
      Kausar Chandra       5.526015

```

Sagnik Ghosh	3.230816
Avaneesh Mahalle	7.106593
Krishna Yadav	4.575089
Viren Jadhav	3.635561
Ketaki Phalle	10.174875
Trisha Kandpal	5.039716
Riya Shirode	1.252007
Kavya Mehta	5.038090
Sumedha Vaidya	2.569386
dtype: float64	

1.14 Popular choice groups

A “Choice group” refers to the candidates voted in a single session.

```
[14]: choice_groups = vote_sessions.value_counts()
```

1.14.1 Most popular choice groups

```
[15]: print(f"Most popular choice groups\n\n{choice_groups.head(5)}")
```

Most popular choice groups

Captain Boy	Captain Girl	Vice Captain Boy	Vice Captain Girl	
Praneel Deshmukh	Gauravi Zade	Avaneesh Mahalle	Ketaki Phalle	24
Aadityaraje Desai	Gauravi Zade	Avaneesh Mahalle	Ketaki Phalle	18
Abhichandra Charke	Gauravi Zade	Avaneesh Mahalle	Ketaki Phalle	14
Aadityaraje Desai	Gauravi Zade	Krishna Yadav	Ketaki Phalle	10
		Kausar Chandra	Ketaki Phalle	9

Name: count, dtype: int64

1.15 Least popular choice groups

```
[16]: print(f"Least popular choice groups \n\n{choice_groups.tail(5)}")
```

Least popular choice groups

Captain Boy	Captain Girl	Vice Captain Boy	Vice Captain Girl
Rachit Srivastava	Kirthika Jayachander	Krishna Yadav	Kavya Mehta
1			
	Naisha Rastogi	Kausar Chandra	Ketaki Phalle
1			
	Tvisha Shah	Avaneesh Mahalle	Ketaki Phalle
1			
		Kausar Chandra	Riya Shirode
1			
		Viren Jadhav	Ketaki Phalle

1

Name: count, dtype: int64

1.16 Average occurrence of a choice group

Average count where all 4 voted candidate were same.

```
[17]: print(f"Average occurrence of a choice group {choice_groups.mean():.2f}")
```

Average occurrence of a choice group 2.81

```
[20]: # Exporting
# !uv run jupyter nbconvert --to html project.ipynb --output index.html
!uv run jupyter nbconvert --to pdf project.ipynb --output index.pdf
```

```
[NbConvertApp] Converting notebook project.ipynb to pdf
[NbConvertApp] Support files will be in index_files/
[NbConvertApp] Making directory ./index_files
[NbConvertApp] Writing 109285 bytes to notebook.tex
[NbConvertApp] Building PDF
[NbConvertApp] Running xelatex 3 times: ['xelatex', 'notebook.tex', '-quiet']
[NbConvertApp] Running bibtex 1 time: ['bibtex', 'notebook']
[NbConvertApp] WARNING | bibtex had problems, most likely because there were no
citations
[NbConvertApp] PDF successfully created
[NbConvertApp] Writing 1199586 bytes to index.pdf
```