

# Sustainability and ESG Data Analytics

## ESG Data Gathering, Analysis, and Visualization

1141ESGDA05

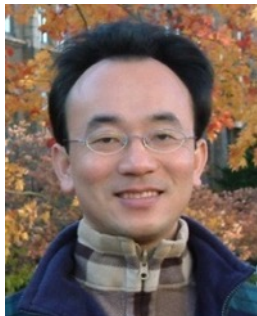
MBA, IM, NTPU (M5265) (Fall 2025)

Wed 2, 3, 4 (9:10-12:00) (B3F17)

 **NVIDIA**  
University Ambassador  
Certified Instructor

 **aws** educate | Cloud  
Ambassador  
2020 Cohort

**Min-Yuh Day, Ph.D,**  
**Professor and Director**

Institute of Information Management, National Taipei University

<https://web.ntpu.edu.tw/~myday>



[https://meet.google.com/  
miy-fbif-max](https://meet.google.com/miy-fbif-max)



# Syllabus

**Week    Date    Subject/Topics**

**1 2025/09/10 Introduction Sustainability and ESG Data Analytics**

**2 2025/09/17 Environmental, Social, and Governance (ESG) in  
Net-Zero Digital Transformation**

**3 2025/09/24 Data Science for Sustainability and ESG**

**4 2025/10/01 Case Study on Sustainability and ESG Data Analytics I**

**5 2025/10/08 Web 3.0 and Big Data Analysis in Fintech, Green and  
Sustainable Finance**

**6 2025/10/15 ESG Data Gathering, Analysis, and Visualization**

# Syllabus

Week	Date	Subject/Topics
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7	2025/10/22	NVIDIA Building RAG Agents with LLMs Part I: LLM Services and AI Foundation Models
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8	2025/10/29	Self-Learning
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9	2025/11/05	Midterm Project Report
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10	2025/11/12	NVIDIA Building RAG Agents with LLMs Part II: Document Loading, Chunking, and Embeddings
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11	2025/11/19	NVIDIA Building RAG Agents with LLMs Part III: Retrieval-Augmented Generation with Vector Stores and RAG Evaluation
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# Syllabus

**Week    Date    Subject/Topics**

**12 2025/11/26 Case Study on Sustainability and ESG Data Analytics II**

**13 2025/12/03 Artificial Intelligence of things (AIoT) in ESG and  
Sustainability Applications**

**14 2025/12/10 Generative AI for ESG Rating and Reporting Generation**

**15 2025/12/17 Final Project Report I**

**16 2025/12/24 Final Project Report II**

# **ESG Data Gathering, Analysis, and Visualization**

# Outline

- **ESG Data Gathering**
- **ESG Data Analysis**
- **ESG Data Visualization**

# ESG Data Challenge: Bridging the Gap

## Traditional: Manual & Siloed

Spreadsheets



Emails & PDFs



Departmental Silos



**Inconsistent, Slow Reporting**



## AI: Automated & Centralized

API & System Integrations



Centralized Data Platform



AI-Powered Analytics



**Auditable, Real-Time Insights**

# ESG Data Gathering

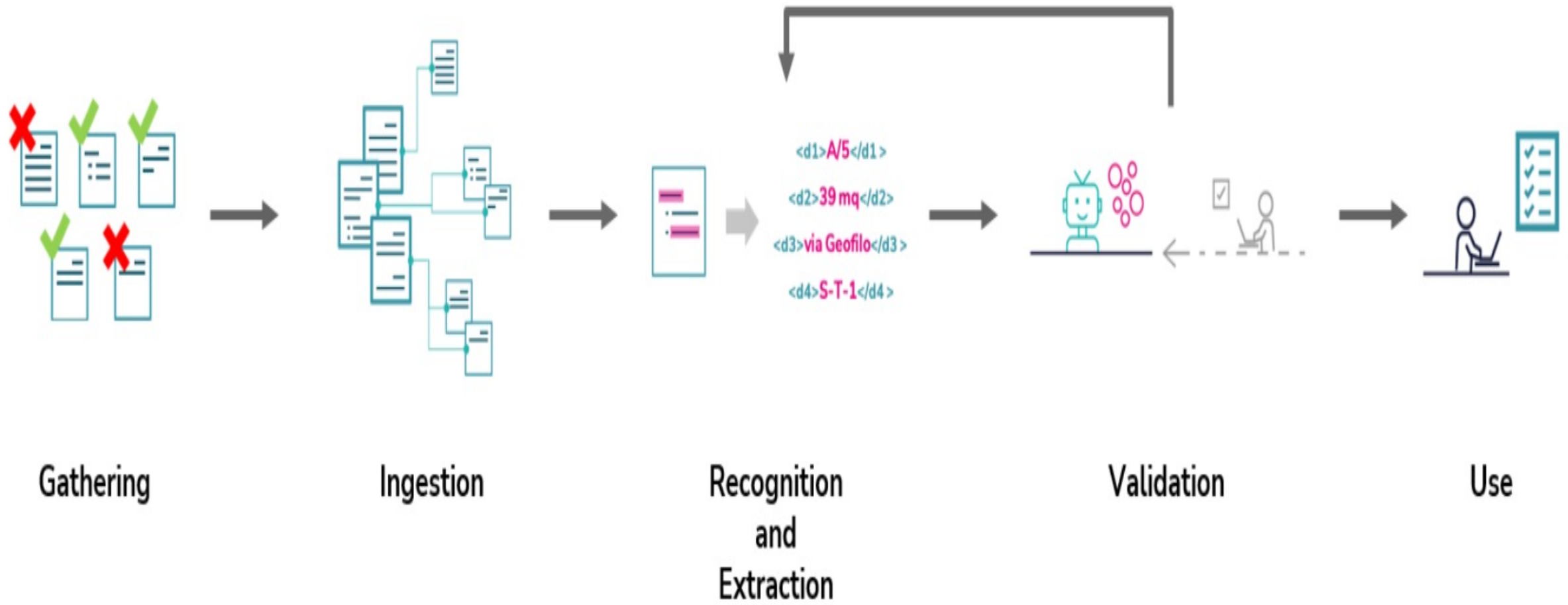
- **Fundamentals of ESG Data**
  - **Defining ESG**
  - **Types of ESG Data**
  - **Key ESG Data Sources**
    - ESG rating providers (MSCI, Sustainalytics, Bloomberg)
    - Corporate sustainability reports
    - Government databases
    - Non-profit and research organizations



# ESG Data Gathering

- **Real-World Example**
  - Analyzing carbon emissions data of a multinational company over time (data source: company reports, CDP)
- **Python Integration**
  - **Python Pandas**
    - Loading and manipulating tabular ESG data
  - **Data Cleaning and Preprocessing**
    - Handling missing values, outliers, and data normalization

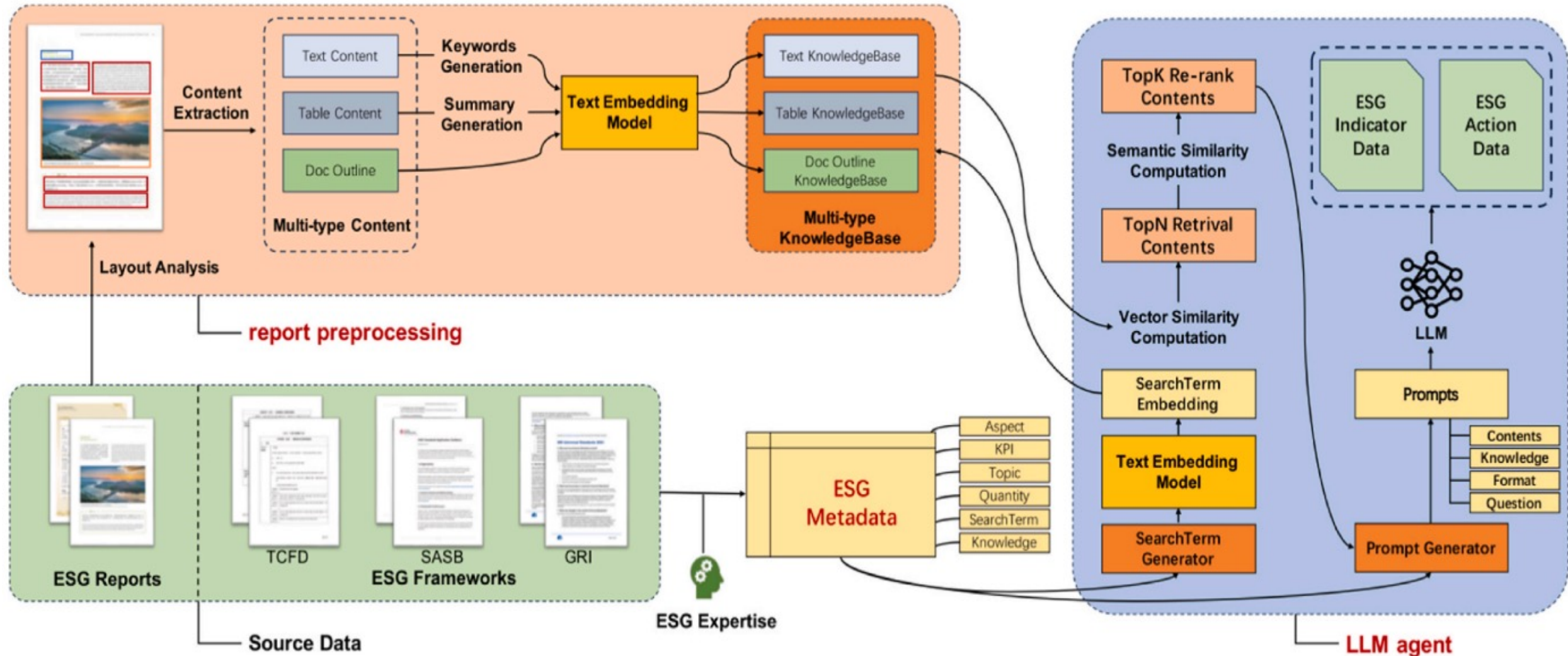
# ESG Data Collection Process



# ESGReveal:

## An LLM-based approach for extracting structured data from ESG reports

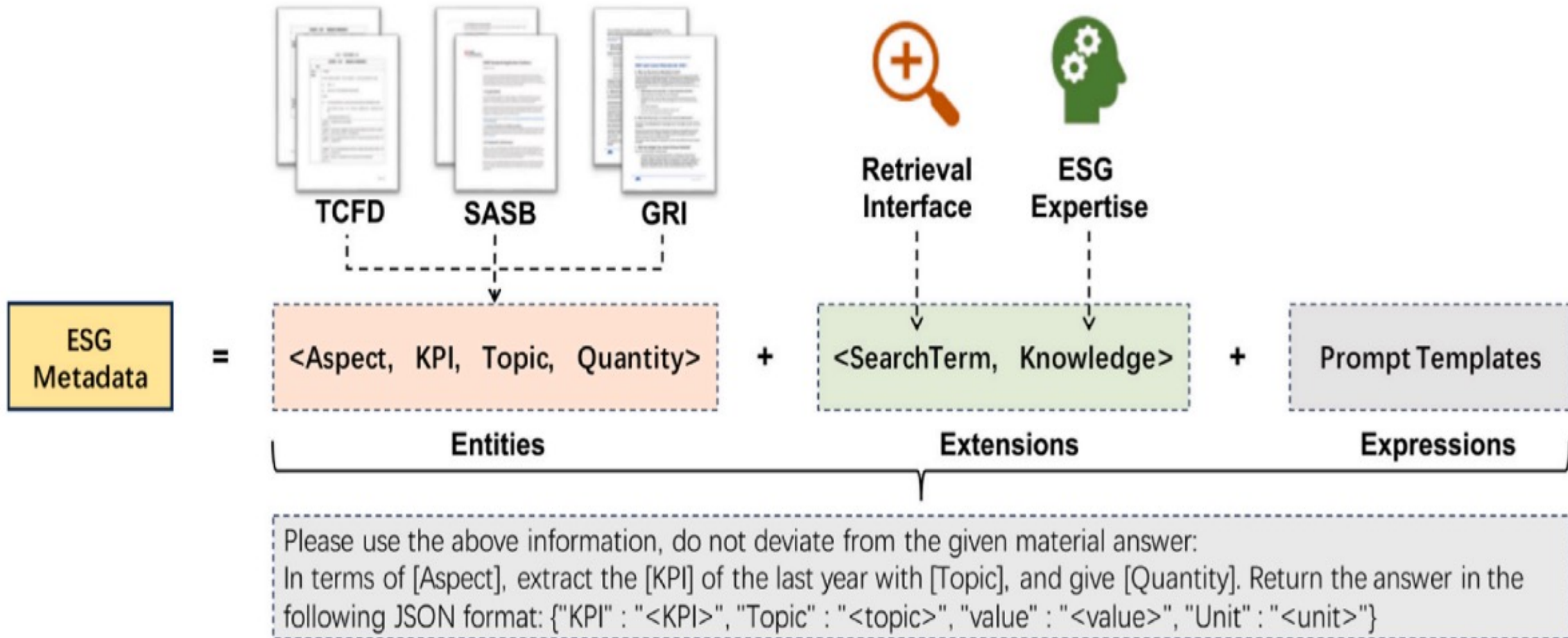
The three-module design (ESG Metadata module, Report Preprocessing module, and LLM Agent module)



Source: Yi Zou, Mengying Shi, Zhongjie Chen, Zhu Deng, ZongXiong Lei, Zihan Zeng, Shiming Yang, Hongxiang Tong, Lei Xiao, and Wenwen Zhou. (2025)

"ESGReveal: An LLM-based approach for extracting structured data from ESG reports." Journal of Cleaner Production 489 (2025): 144572.

# ESGReveal ESG Metadata Module: Entities, Extensions, and Expressions



# ESGReveal ESG Metadata:

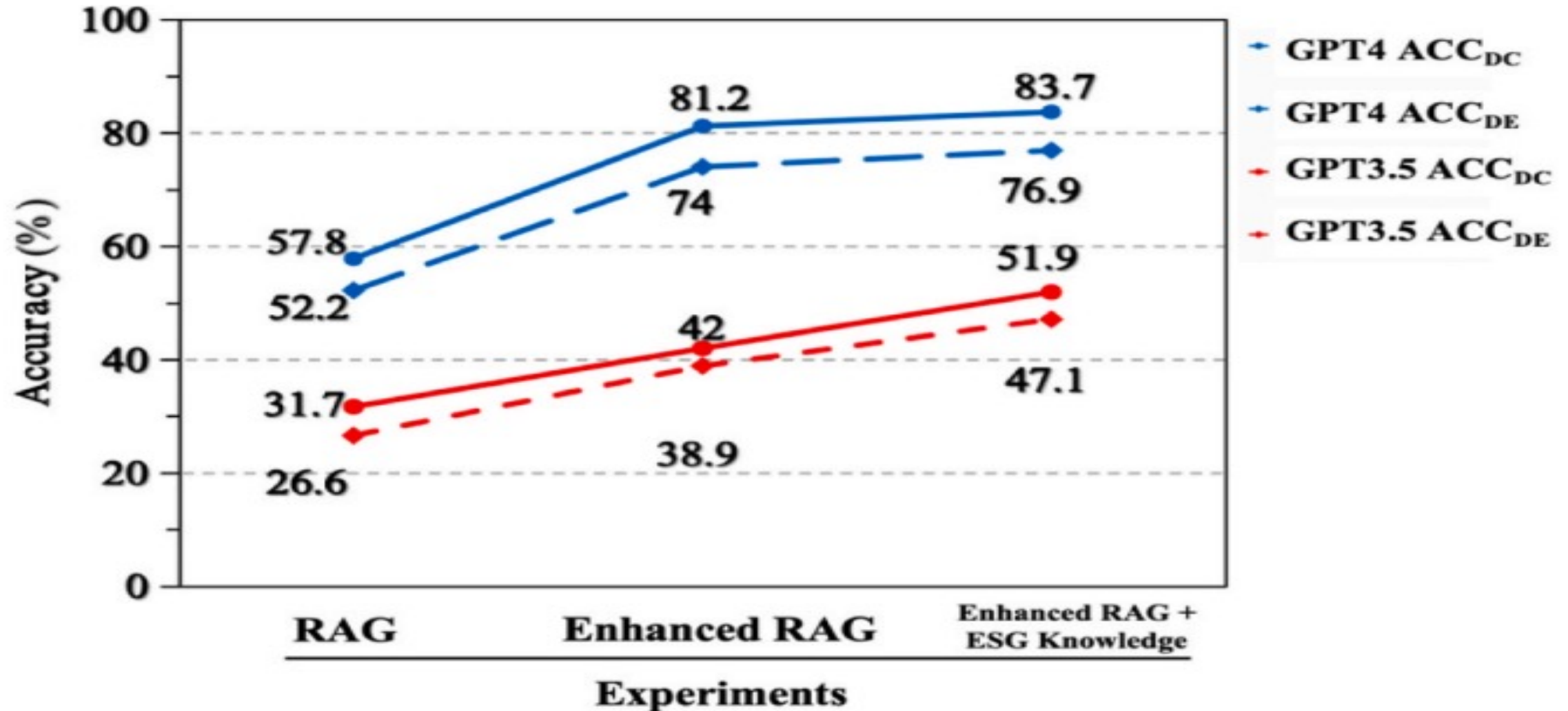
## An example of the main contents of one KPI

Aspect	KPI	Topic	Quantity	Search Term	Knowledge
A1. Emissions	Total waste produced (in tonnes) and, where appropriate, intensity (e.g., per unit of production volume, per facility)	[Non-hazardous Waste, Hazardous Waste]	[Absolute Values]	Construction waste, building rubbish, soil, rubble, organic waste, recyclable fertilizer, paper, plastic bottles, wood, household garbage	Hazardous waste typically includes the following categories: toxic chemical waste, electronic waste, hazardous gas waste, dangerous waste.



# Ablation study on ESGReveal

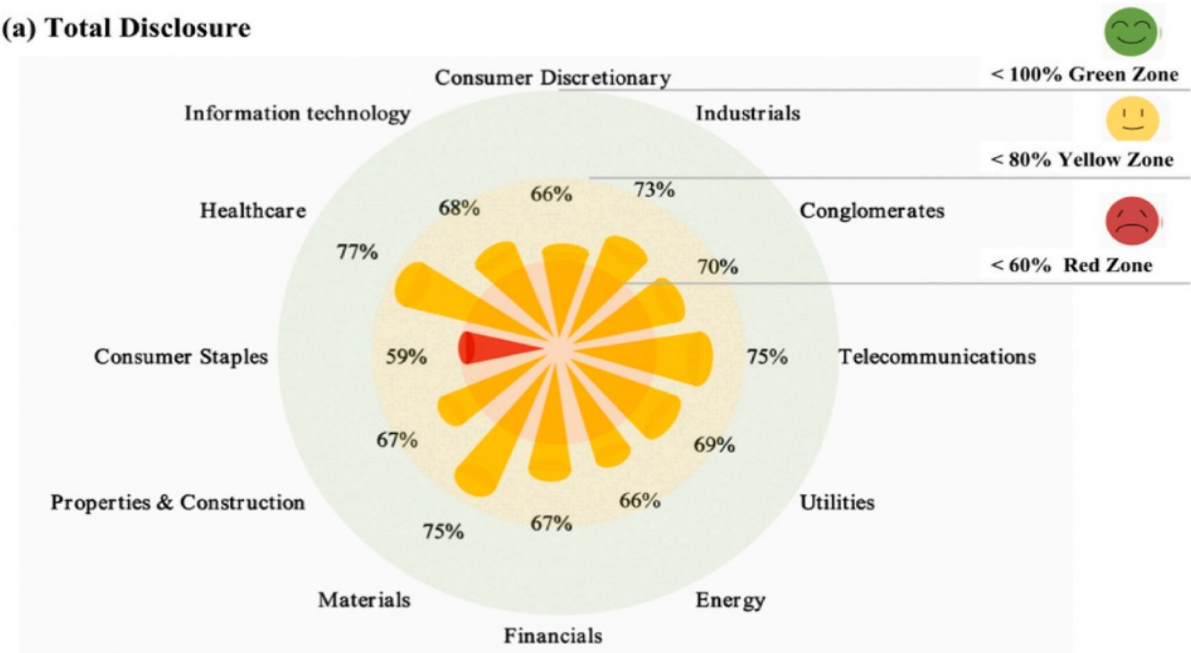
performance of data extraction accuracy (AccDE) and disclosure coverage accuracy (AccDC) by integrating different modules



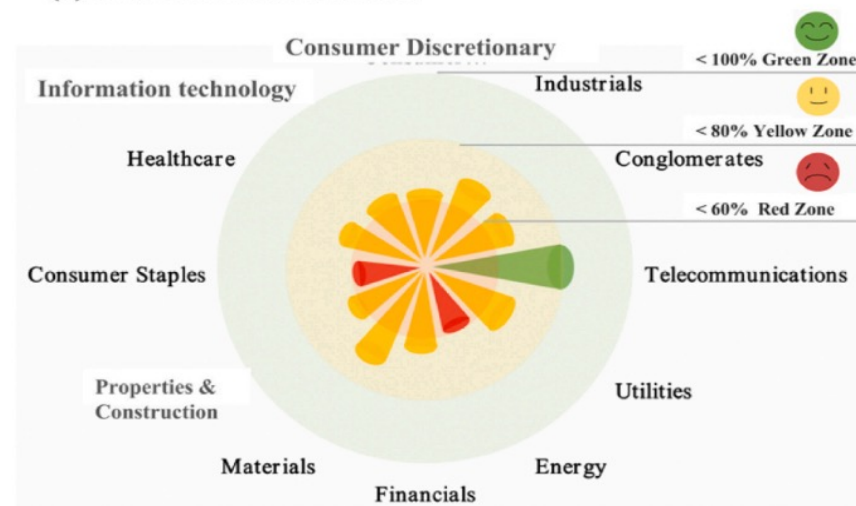
# ESGReveal Disclosure levels by sector:

- (a) overall,
- (b) environmental  
issues
- (c) social issues.

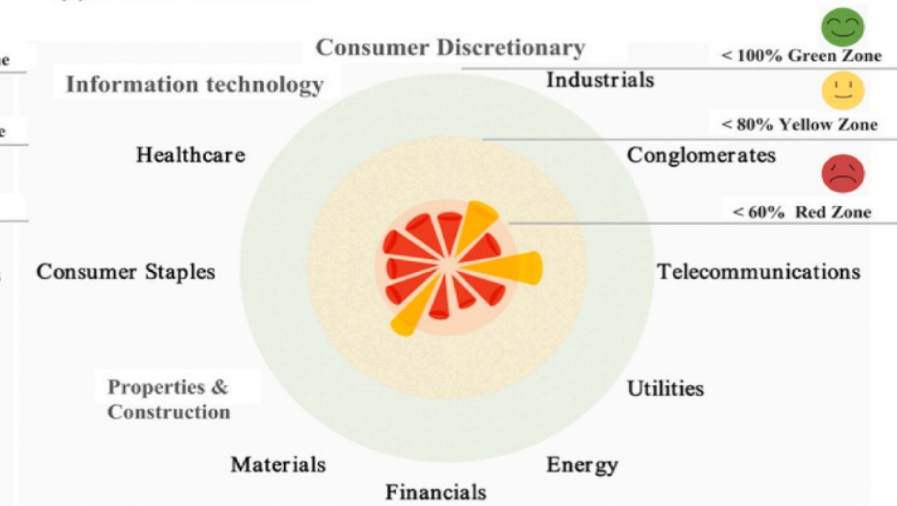
(a) Total Disclosure



(b) Environmental Disclosure



(c) Social Disclosure



# SusGen-GPT:

## A Data-Centric LLM for Financial NLP and Sustainability Report Generation

### Financial NLP Tasks

Relation Extraction

Financial Question Answering

Headline Classification

Financial Table Question Answering

Name Entity Recognition

Text Summarization

Sentiment Analysis

ESG Question Answering

Sustainability Report Generation

Prompt

### SusGen System



SusGen-GPT

Response

Answer

RAG



Annual Report



TCFD Report



# Analyst-driven vs. AI-driven ESG

## Analyst-driven ESG research

Derives ratings in a structured data model



## Sustainalytics

*Analyst role at the end of the process allows subjectivity to color results*

## AI-driven ESG research

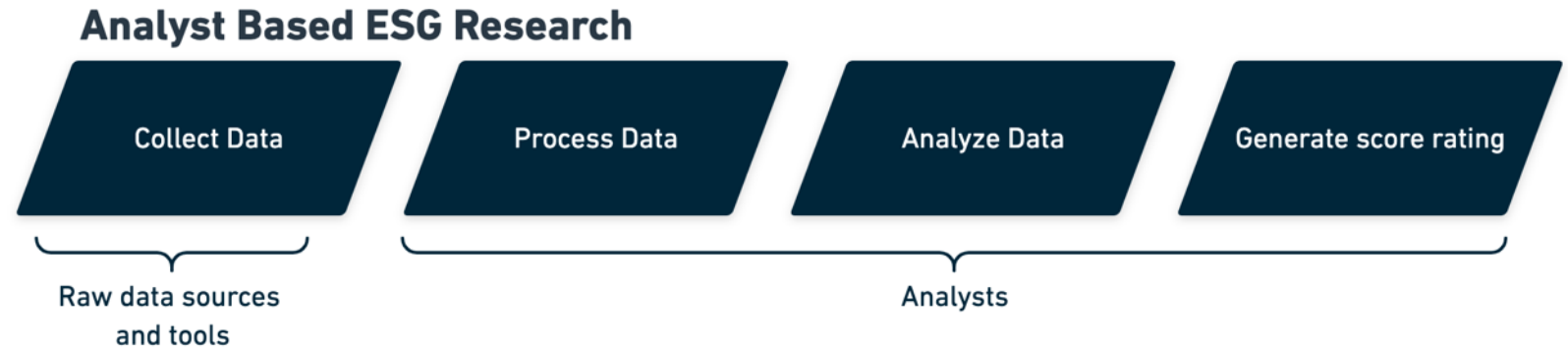
Derives signals from unstructured data



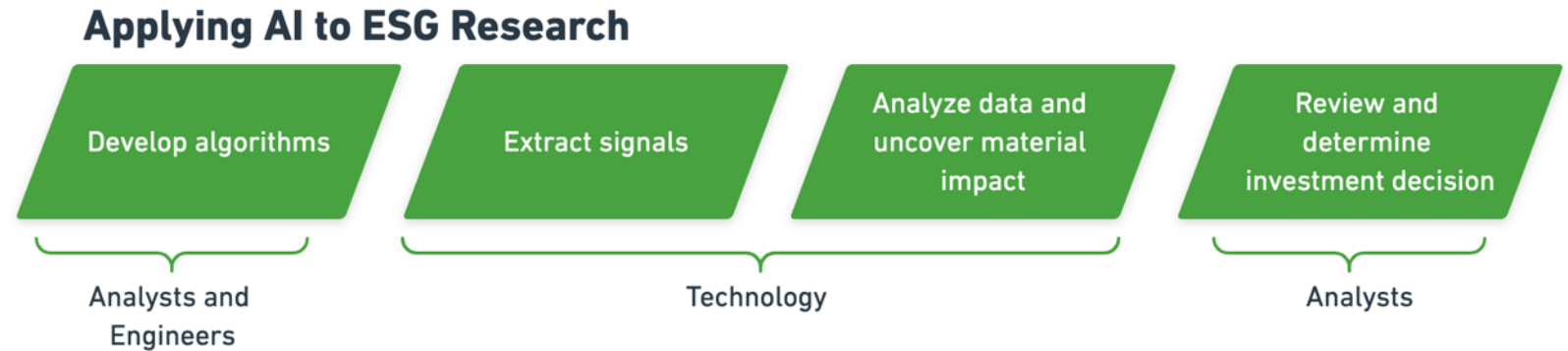
## Truvalue Labs

*Analyst expertise at the beginning of the process produces consistent results*

# Analyst based ESG Research

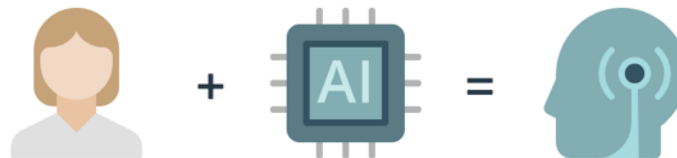


# AI based ESG Research



**It would take an analyst over 5 years to do what our AI can in 1 week**

Combining analysts with AI creates gives you the full picture



**ESG ANALYTICS**  
Invest where it matters.

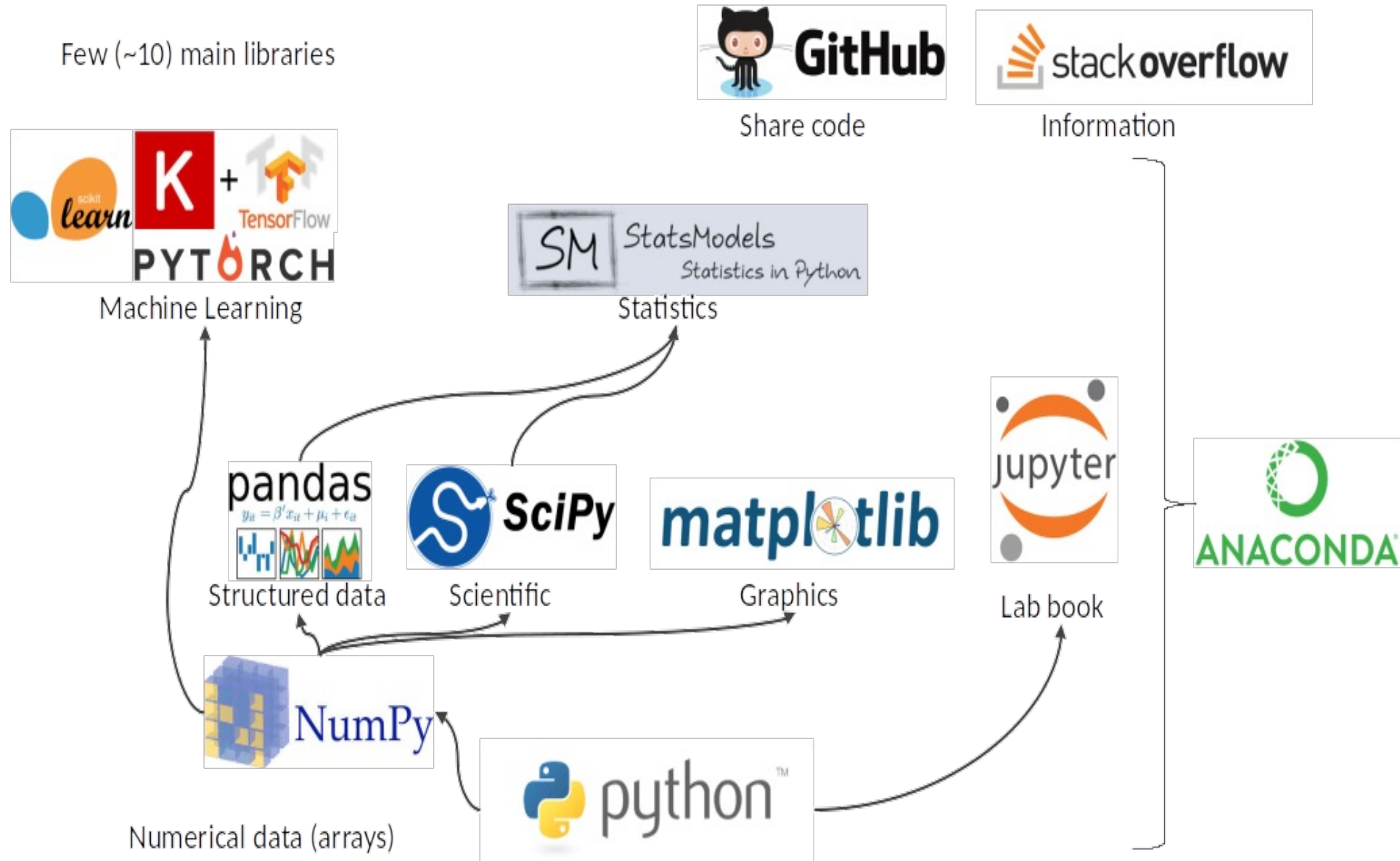
# ESG Data Analysis

- **ESG Metrics and Frameworks**
  - Popular frameworks (GRI, SASB, TCFD)
  - Calculating key ESG ratios
    - (e.g., debt-to-equity, employee turnover rate, energy efficiency metrics)
- **Statistical Analysis of ESG Data**
- **Time Series Analysis**

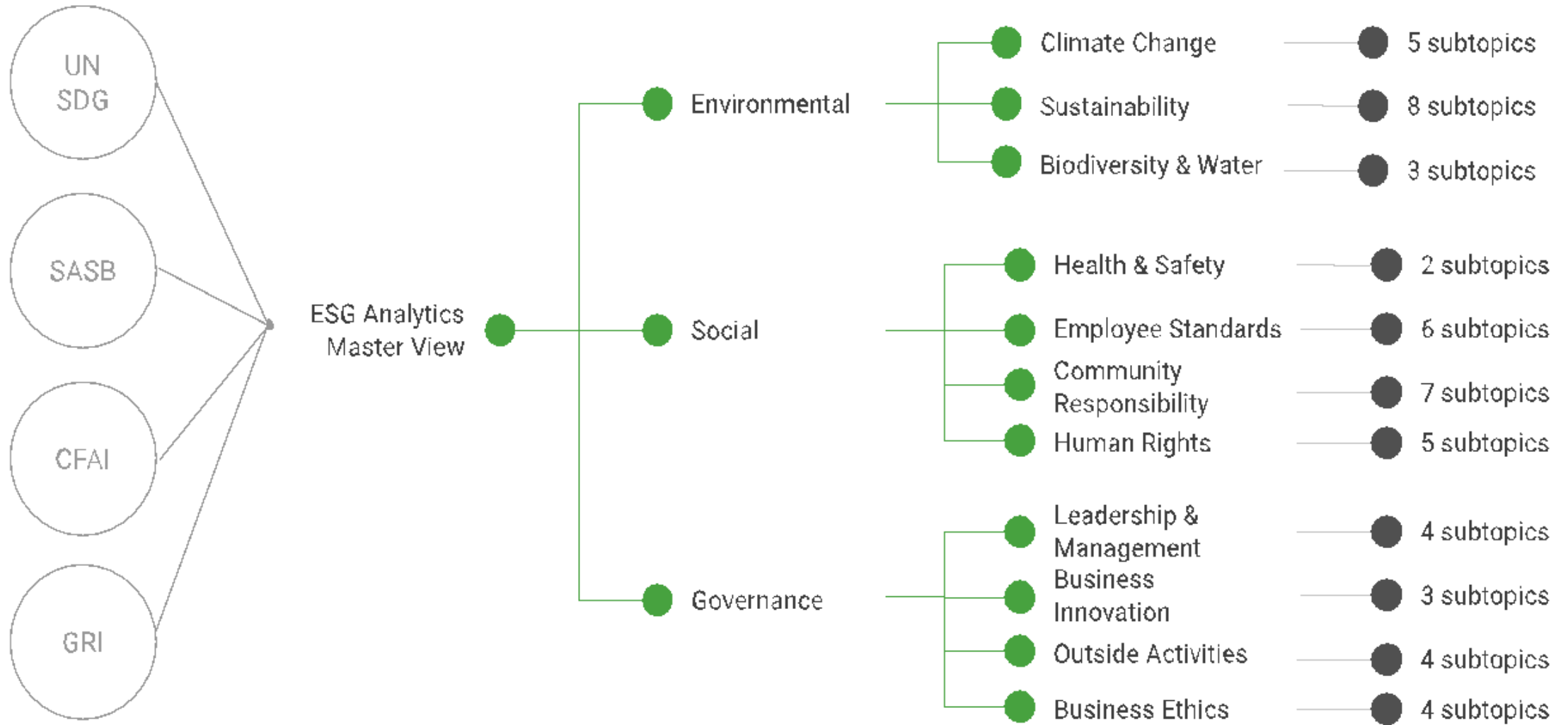
# ESG Data Analysis

- **Real-World Example**
  - **Examining the correlation between gender diversity on corporate boards and profitability**
    - (data source: Bloomberg, MSCI)
- **Python Integration**
  - **Statistical Libraries**
    - Using SciPy and NumPy for calculations
  - **Time Series Analysis**
    - Utilizing Pandas for date/time manipulation and analysis

# Python Ecosystem for Data Science



# ESG Analytics: NLP Taxonomy



# ESG Data Visualization

- **Principles of Effective Data Visualization**
- **ESG Dashboards**
- **Interactive Visualizations**
- **Real-World Example: Visualizing a company's ESG performance against industry benchmarks using a dashboard.**

# ESG Data Visualization

- **Real-World Example**
  - Visualizing a company's ESG performance against industry benchmarks using a dashboard.
- **Python Integration**
  - **Visualization Libraries**
    - Matplotlib, Seaborn, or Plotly
  - **Dashboarding Tools**
    - Streamlit or Dash for creating simple interactive dashboards



# Fundamentals of ESG Data

- **ESG Data Gathering, Analysis, and Visualization: Practical Techniques and Python Applications**
- **What ESG data is**
  - where to find it
  - its importance in the world of sustainability

# Environmental, Social, and Governance (ESG)

- It's a framework used to evaluate the **sustainability** and **ethical impact** of **companies** and **investments**.
- **Environmental**
  - company's interaction with the natural world.
- **Social**
  - company's relationships with people – its workers, customers, and communities.
- **Governance**
  - how a company is led and managed.

# Defining ESG

- **E – Environmental**
  - carbon emissions, pollution, resource use, biodiversity impact.
- **S – Social**
  - Labor practices, human rights, diversity & inclusion, community impact.
- **G – Governance**
  - Board structure, executive compensation, ethics, transparency, anti-corruption.

# Importance of ESG

- **Growing investor demand**
  - Investors are increasingly looking for companies aligned with their values on sustainability.
- **Risk mitigation**
  - Strong ESG performance can help companies avoid potential scandals and reputational damage.
- **Long-term value creation**
  - ESG factors are linked to better resource management, innovation, and improved financial performance over time.

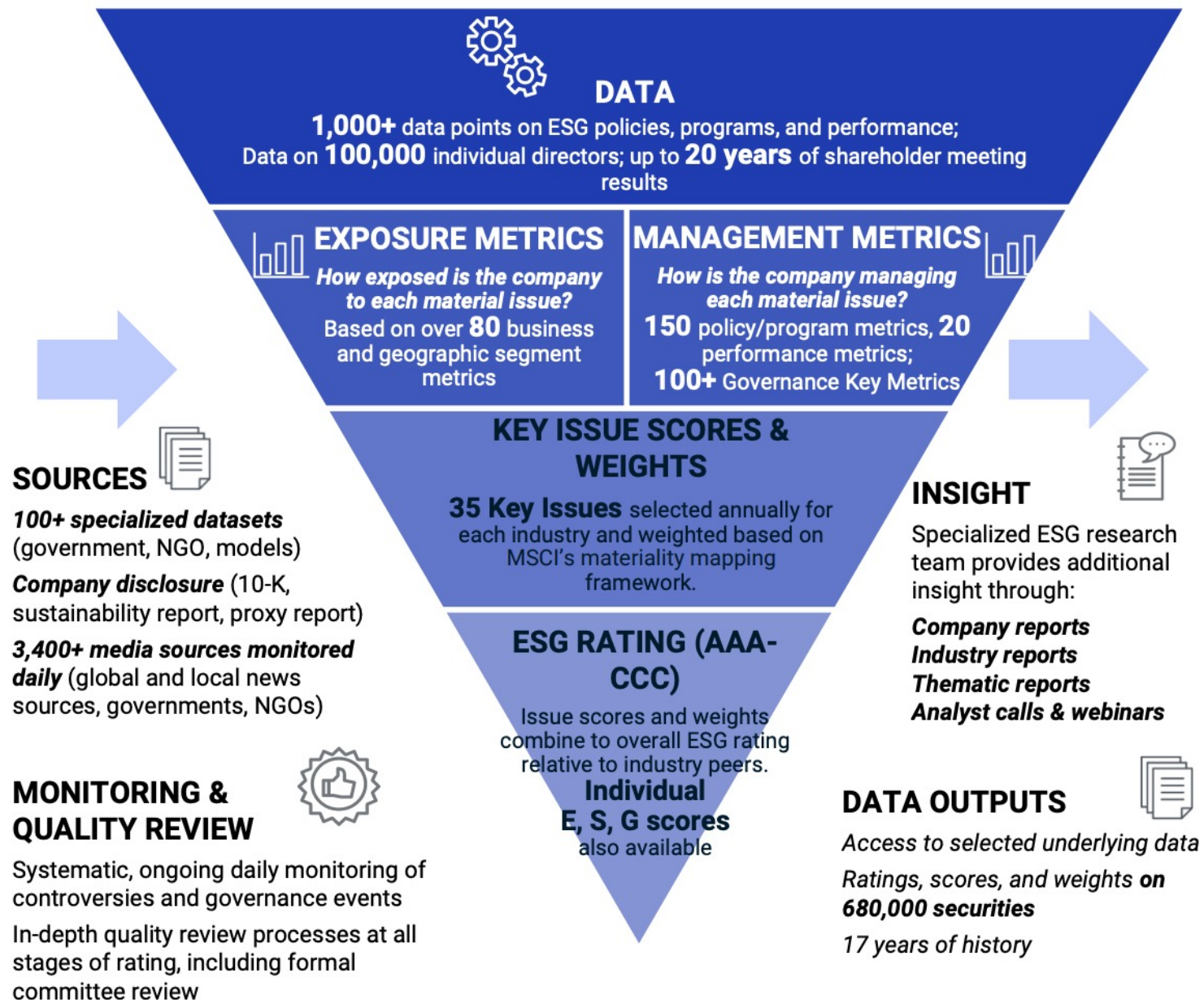
# Types of ESG Data

- **Qualitative**
  - **Narrative information, such as policies, statements, news articles**
- **Quantitative**
  - **Numerical data, like emissions figures, employee demographics, board compensation**
- **Structured**
  - **Data that fits neatly into tables and databases**
- **Unstructured**
  - **Data like news articles, social media posts, or company reports that need further processing**

# Key ESG Data Sources

- **ESG Rating Providers**
  - Companies specializing in ESG assessments and data
- **Corporate Sustainability Reports**
  - Company-issued reports on their ESG activities
- **Government Databases**
  - Public data on environmental metrics, labor statistics, etc.
- **Non-profit & Research Orgs**
  - Issue-specific data on topics like human rights or climate change

# MSCI ESG Rating Framework

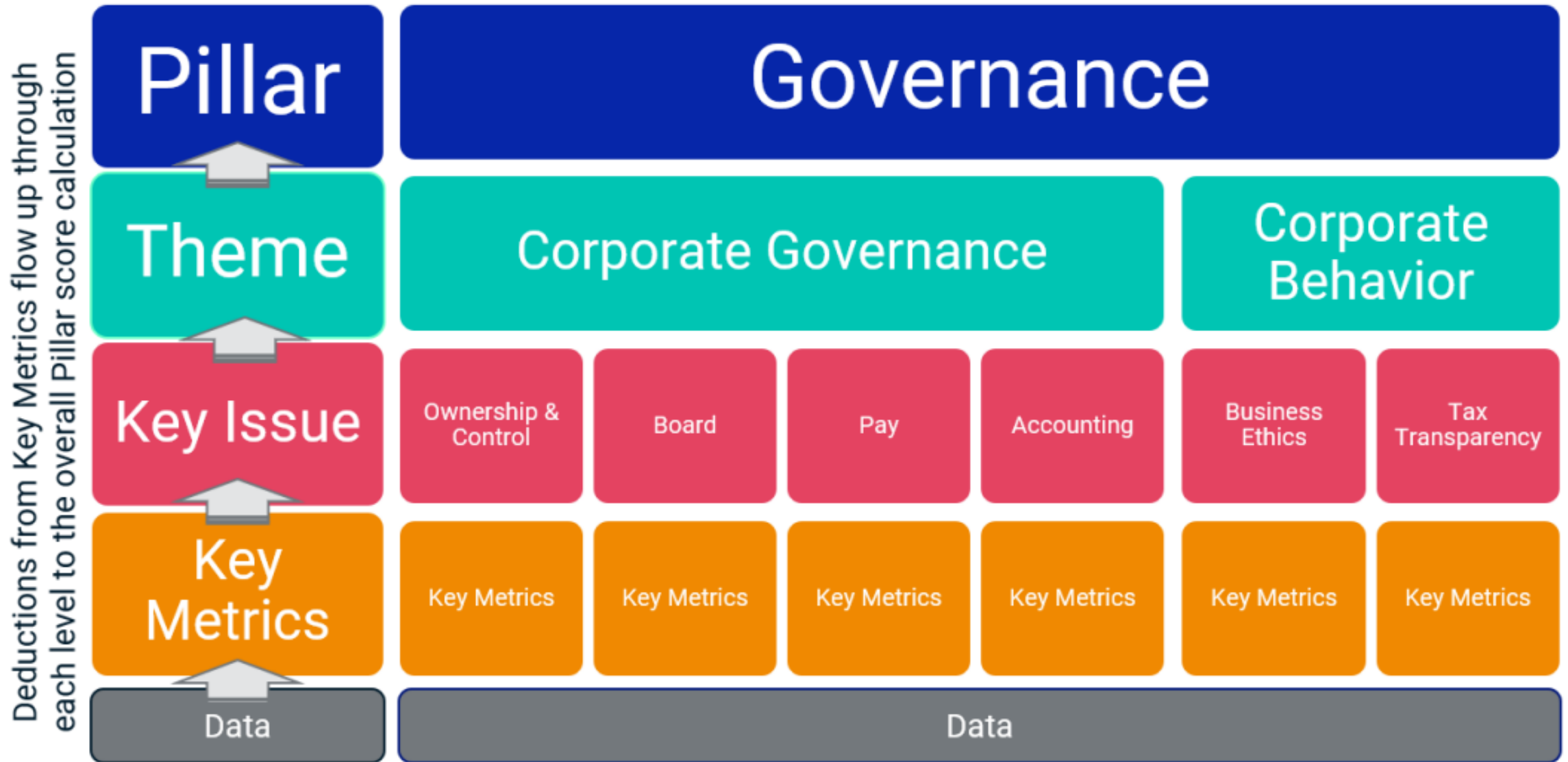


# MSCI ESG Key Issue Hierarchy

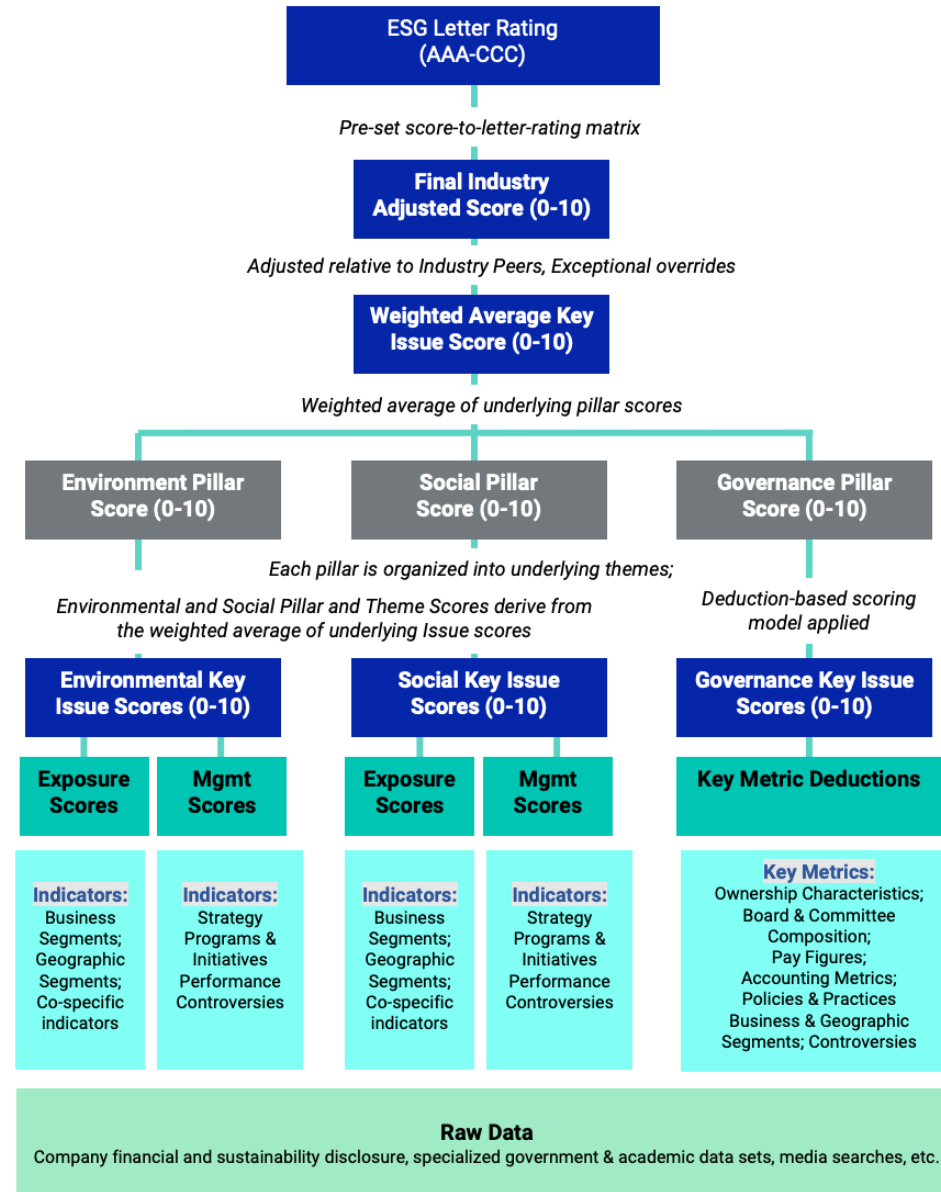
3 Pillars	10 Themes	35 ESG Key Issues	
<b>Environment</b>	<b>Climate Change</b>	Carbon Emissions Product Carbon Footprint	Financing Environmental Impact Climate Change Vulnerability
	<b>Natural Capital</b>	Water Stress Biodiversity & Land Use	Raw Material Sourcing
	<b>Pollution &amp; Waste</b>	Toxic Emissions & Waste Packaging Material & Waste	Electronic Waste
	<b>Environmental Opportunities</b>	Opportunities in Clean Tech Opportunities in Green Building	Opportunities in Renewable Energy
<b>Social</b>	<b>Human Capital</b>	Labor Management Health & Safety	Human Capital Development Supply Chain Labor Standards
	<b>Product Liability</b>	Product Safety & Quality Chemical Safety Consumer Financial Protection	Privacy & Data Security Responsible Investment Health & Demographic Risk
	<b>Stakeholder Opposition</b>	Controversial Sourcing Community Relations	
	<b>Social Opportunities</b>	Access to Communications Access to Finance	Access to Health Care Opportunities in Nutrition & Health
<b>Governance</b>	<b>Corporate Governance</b>	Ownership & Control Board	Pay Accounting
	<b>Corporate Behavior</b>	Business Ethics Tax Transparency	



# MSCI Governance Model Structure



# MSCI Hierarchy of ESG Scores



# DJSI S&P Global ESG Score

**8,000**

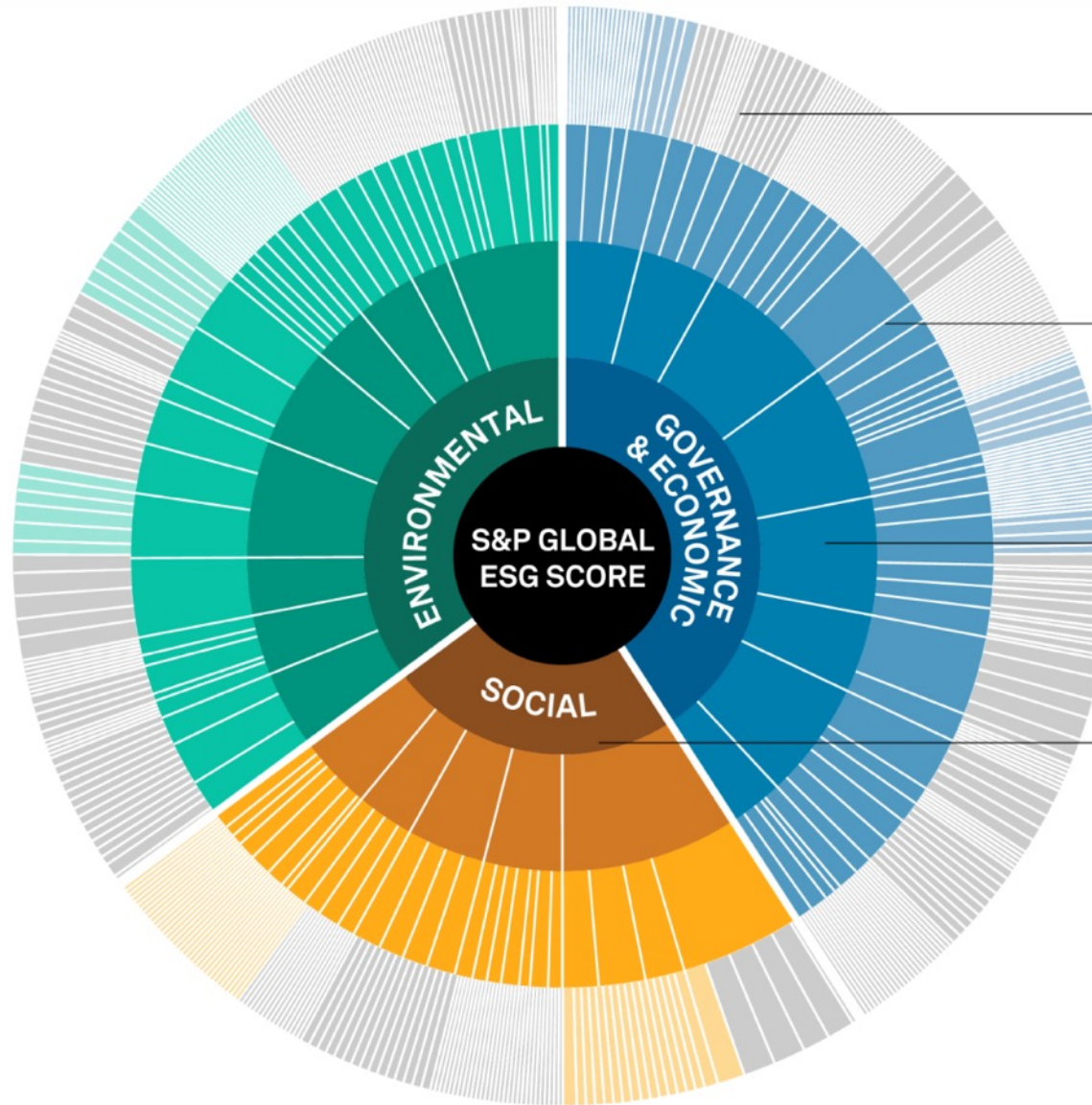
Companies

**90%**

Global market  
capitalization

**340,000+**

Current Research Universe  
and Active Securities



Approx.  
**1,000**  
Datapoints

**Assessed values, text,  
checkboxes, documents**

Sources: Web-based questionnaire  
and company documents

**130+**  
Questions

**Weighted  
data point scores**

Up to 50% industry-specific

Ave.  
**30+**  
Criteria scores

**Weighted  
question scores**

61 industry specific approaches,  
with tailored questions, criteria  
and related weightings

**3**  
Dimension scores

**Weighted  
criteria scores**

Adjusted for corporate ESG  
controversies where applicable

**1**

**S&P Global  
ESG Score**

**Sum of weighted  
dimension scores**

# FTSE Russell ESG Ratings



# ESG Data Analysis

- **ESG Metrics and Frameworks**
- **Key ESG Ratios**
- **Statistical Analysis of ESG Data**
- **Time Series Analysis**

# ESG Metrics and Frameworks

- **Metrics:**
  - **Specific ESG measures**  
(e.g., carbon intensity, employee satisfaction, percentage of independent directors).
- **Frameworks:**
  - **Standards for reporting and interpreting ESG data**  
(GRI, SASB, TCFD).



# Key ESG Ratios

## Financial Ratios with an ESG Twist

- **Debt-to-equity**
  - Indicates financial risk, but also potential social or environmental liabilities.
- **Employee turnover rate**
  - Measure of employee satisfaction and human capital management.
- **Energy efficiency metrics**
  - Energy use per unit of output (shows resource optimization)

# Statistical Analysis of ESG Data

## Numbers Tell a Story

- **Descriptive statistics**
  - Summarize data (mean, median, standard deviation)
- **Correlation**
  - Measures the association between two ESG variables
- **Regression**
  - Models the relationship between ESG factors and financial performance



# Time Series Analysis

## The Power of Tracking Over Time

- **Detecting trends**
  - Are a company's emissions decreasing?
  - Is board diversity improving?
- **Identifying patterns**
  - Understanding seasonal or cyclical fluctuations in ESG data.

# Real-World Example

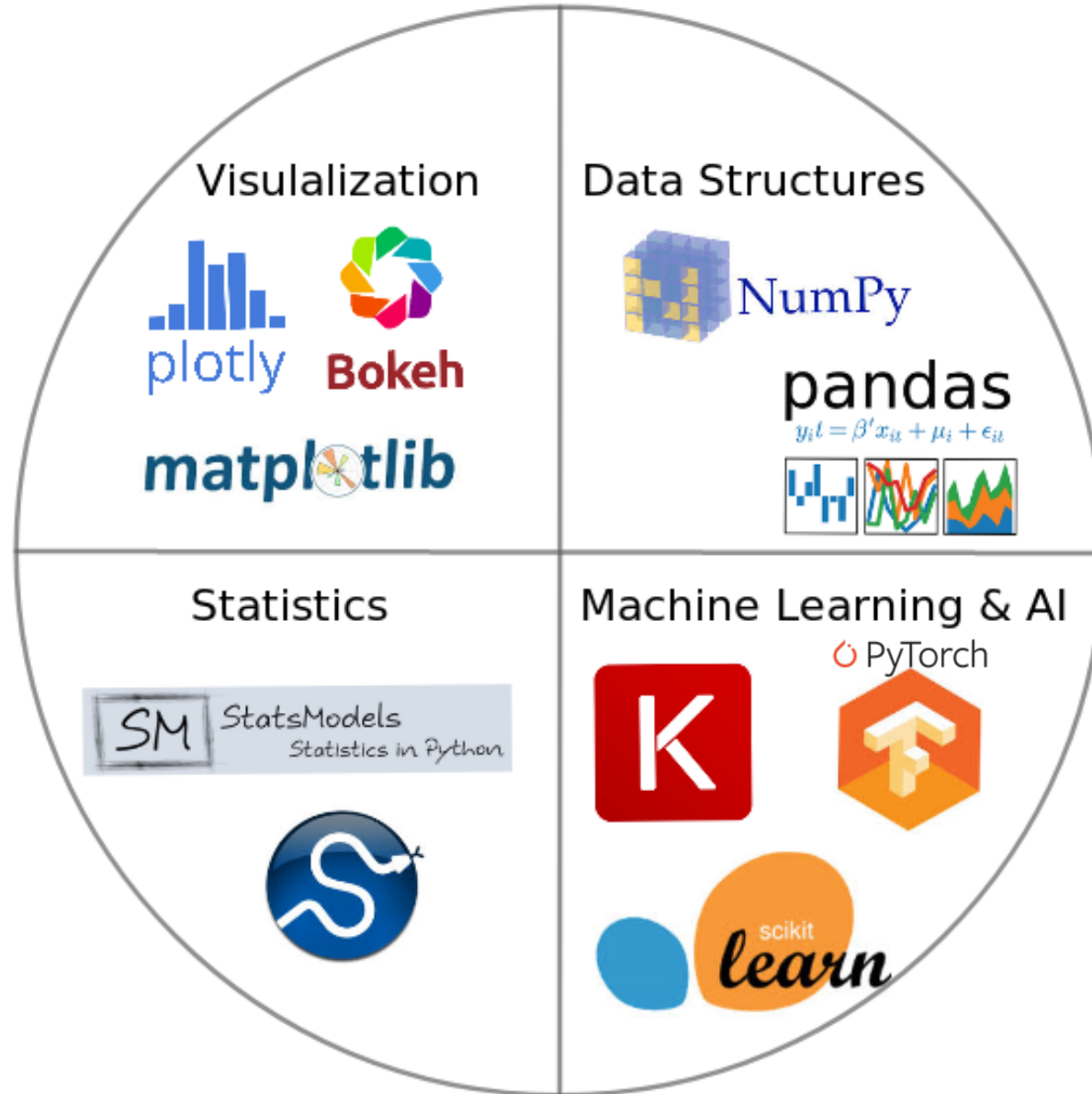
## Gender Diversity and Profitability: A Correlation?

- **Data Source:**
  - Bloomberg, MSCI
- **Metrics:**
  - Percentage of women on boards,
  - Return on Equity (or similar profitability metric)

# Python: Your ESG Analysis Toolbox

- **SciPy & NumPy**
  - Powerful libraries for calculations and statistical analysis
- **Pandas**
  - Ideal for working with time series data and tabular ESG data

# Python Ecosystem for Data Science



# ESG Data Visualization

- **Principles of Effective Data Visualization**
- **Chart Types for ESG**
- **Storytelling with Data**
- **ESG Dashboards**

# **Data Analytics and Visualization with Python**

# Data Analytics and Visualization with Python

- **NumPy**
  - Numerical Python N-dimensional array
- **Pandas**
  - Data Analytics
- **Matplotlib**
  - Basic Data Visualization
- **Seaborn**
  - Advanced Visualization

# Python Data Analysis and Visualization





# Principles of Effective Data Visualization

## Visualizing ESG: Clarity and Impact

- **Clarity**
  - Choose chart types that clearly communicate the message
- **Simplicity**
  - Avoid visual clutter, highlight key takeaways
- **Aesthetics**
  - Use color and design purposefully

# Chart Types for ESG

- **Bar Charts**
  - **Comparing values across categories**  
(emissions of different companies)
- **Line Charts**
  - **Showing change over time**  
(a company's energy use over years)
- **Scatter Plots**
  - **Examining correlations**  
(diversity vs. employee satisfaction)
- **Pie Charts**
  - **Part-to-whole relationships**  
(breakdown of waste types)

# Storytelling with Data

## More Than Just Numbers

- **Narrative**
  - **Frame your visualizations with context and key takeaways**
- **Annotations**
  - **Use text labels and callouts directly on charts to highlight insights**
- **Visual Hierarchy**
  - **Guide the viewer's eye to the most important information**

# **ESG Dashboards**

## **Your ESG Mission Control**

- **Key Metrics**
  - **Track the most important indicators in one place**
- **Comparisons**
  - **Visualize performance against targets and benchmarks**
- **Interactivity**
  - **Allow users to filter data by time, company, or metric**

# ESG Data Visualization

## Company vs. Industry

### An ESG Dashboard



# Python: Your Visualization Powerhouse

- **Matplotlib & Seaborn**
  - Foundational libraries for static plots
- **Plotly**
  - For highly interactive visualizations
- **Streamlit or Dash**
  - Tools for creating dashboard web applications

# ESG Data Analysis and Visualization

```
# ESG Data Analysis and Visualization
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from datetime import datetime
import random

# Generate synthetic data
np.random.seed(0)
data = {
    'company': ['Company A', 'Company B', 'Company C', 'Company D', 'Company E'],
    'emissions': np.random.randint(10000, 50000, 5),
    'diversity': np.random.uniform(0.2, 0.9, 5),
    'employee_satisfaction': np.random.uniform(60, 90, 5),
    'waste_type': ['Plastic', 'Organic', 'Electronic', 'Metal', 'Other'],
    'waste_amount': np.random.randint(100, 500, 5)
}
df = pd.DataFrame(data)
```

# ESG Data Analysis and Visualization

```
# Separate DataFrame for time series and correlation
time_series_data = pd.DataFrame({
    'year': np.repeat(np.arange(2018, 2023), 5),
    'company': np.tile(['Company A', 'Company B', 'Company C', 'Company D', 'Company E'], 5),
    'energy_use': np.random.randint(1000, 5000, 25)
})

# Simulating correlation data with a slight positive trend
diversity = np.linspace(0.2, 0.9, 100)
np.random.shuffle(diversity)
employee_satisfaction = 60 + (diversity - 0.2) * 150
employee_satisfaction += np.random.normal(0, 5, 100)

correlation_data = pd.DataFrame({
    'diversity': diversity,
    'employee_satisfaction': employee_satisfaction
})
```



# ESG Data Analysis and Visualization

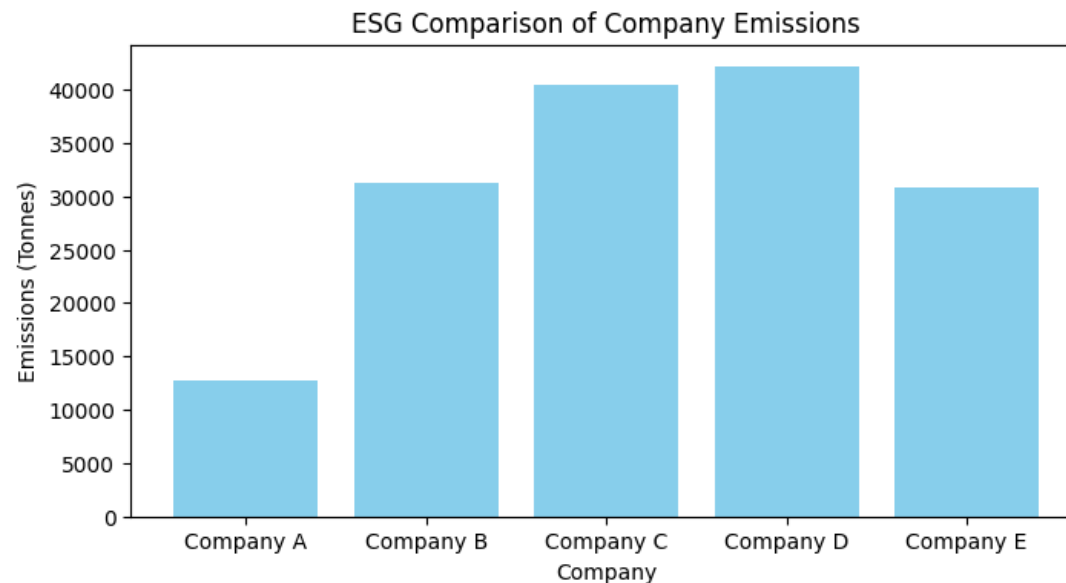
```
# Save DataFrame to CSV
df.to_csv('ESG_Dataset.csv', index=False)

# Calculate statistics for each company
statistics = df.describe()
statistics = statistics.applymap(lambda x: format(x, '.4f'))
print(statistics)
statistics.to_csv('Company_ESG_Statistics.csv')
```

	emissions	diversity	employee_satisfaction	waste_amount
count	5.0000	5.0000	5.0000	5.0000
mean	31447.6000	0.4085	77.1504	345.6000
std	11667.5748	0.1099	7.0841	94.8093
min	12732.0000	0.2397	70.1219	215.0000
25%	30757.0000	0.3909	71.7835	297.0000
50%	31243.0000	0.4083	74.3993	343.0000
75%	40403.0000	0.4691	84.3651	435.0000
max	42103.0000	0.5344	85.0824	438.0000

# ESG Data Analysis and Visualization

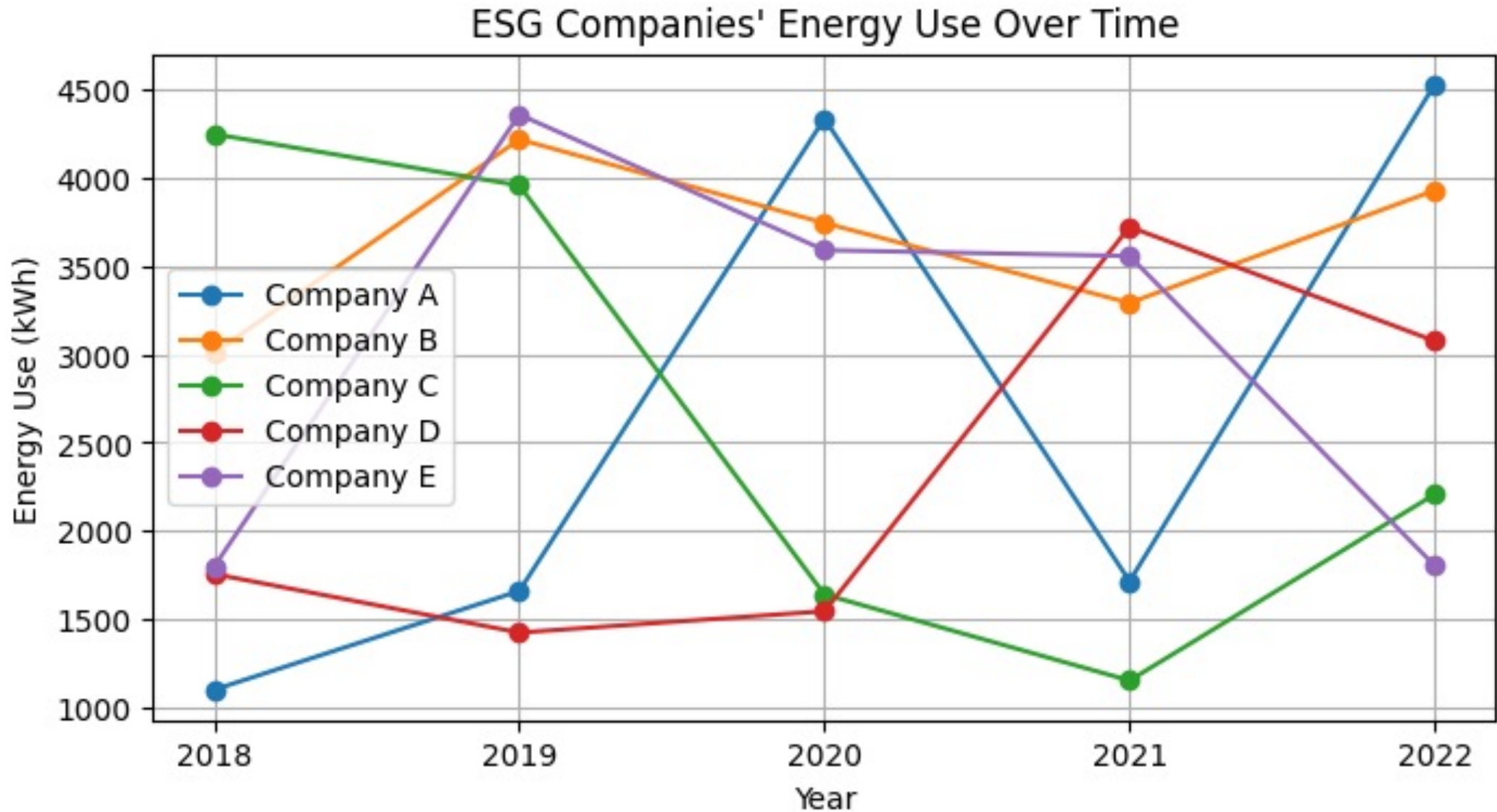
```
# Create visualizations and save them at 300 dpi
# Bar Chart for Emissions
plt.figure(figsize=(8, 4))
plt.bar(df['company'], df['emissions'], color='skyblue')
plt.xlabel('Company')
plt.ylabel('Emissions (Tonnes)')
plt.title('ESG Comparison of Company Emissions')
plt.show()
plt.savefig('ESG Company_Emissions.jpg', format='jpg', dpi=300)
```



# ESG Data Analysis and Visualization

```
# Line Chart for Energy Use
plt.figure(figsize=(8, 4))
for company in time_series_data['company'].unique():
    company_data = time_series_data[time_series_data['company'] ==
    company]
    company_data = company_data.sort_values(by='year')
    plt.plot(company_data['year'], company_data['energy_use'],
    marker='o', linestyle='-', label=company)
plt.xlabel('Year')
plt.ylabel('Energy Use (kWh)')
plt.title("ESG Companies' Energy Use Over Time")
plt.xticks(company_data['year'].unique()) # Ensuring only whole years
are marked
plt.legend()
plt.grid(True)
plt.show()
```

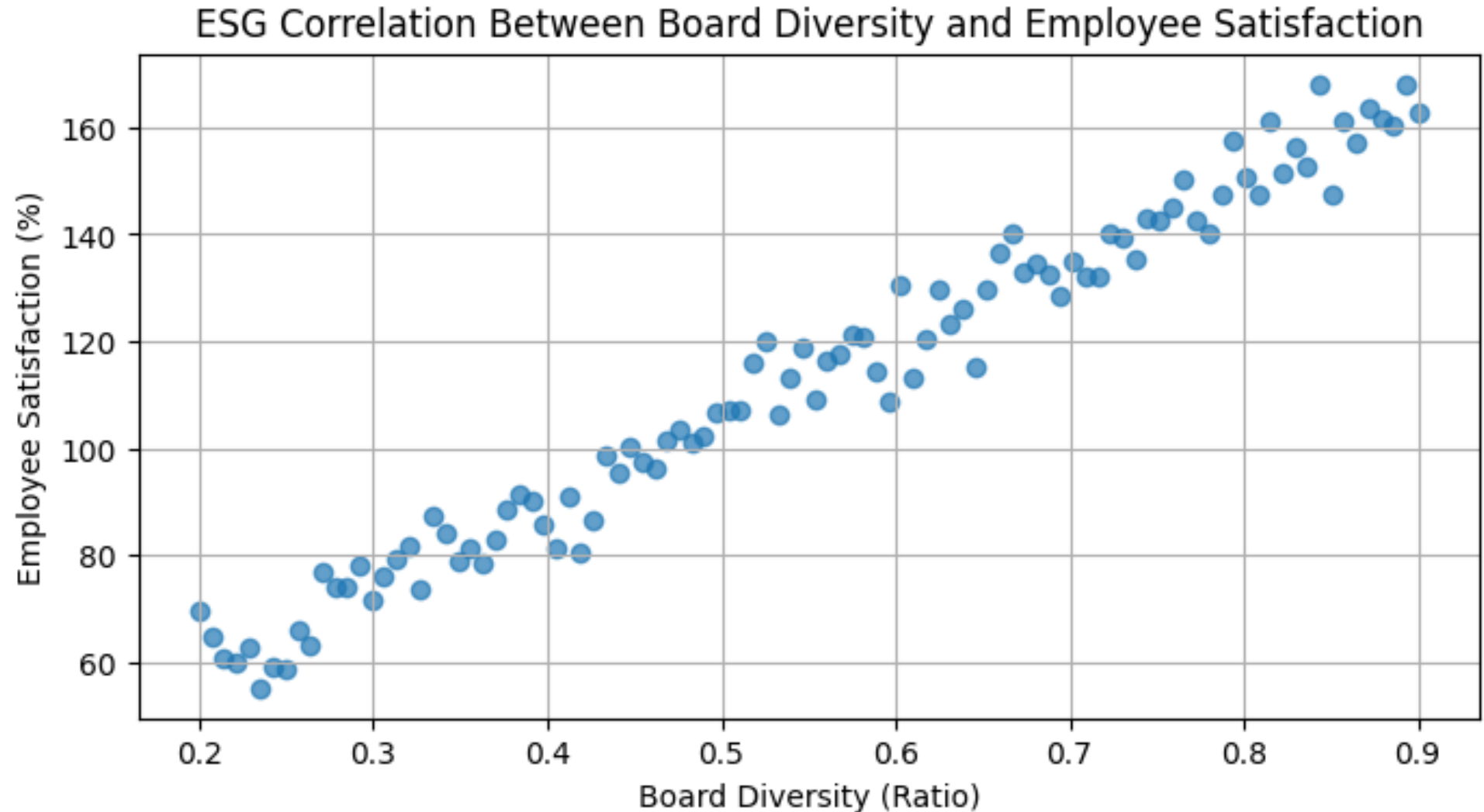
# ESG Data Analysis and Visualization



# ESG Data Analysis and Visualization

```
# Scatter Plot for Diversity vs. Satisfaction
plt.figure(figsize=(8, 4))
plt.scatter(correlation_data['diversity'],
            correlation_data['employee_satisfaction'], alpha=0.7)
plt.xlabel('Board Diversity (Ratio)')
plt.ylabel('Employee Satisfaction (%)')
plt.title('ESG Correlation Between Board Diversity and Employee
Satisfaction')
plt.grid(True)
plt.show()
plt.savefig('ESG_Diversity_vs_Satisfaction.jpg', format='jpg', dpi=300)
```

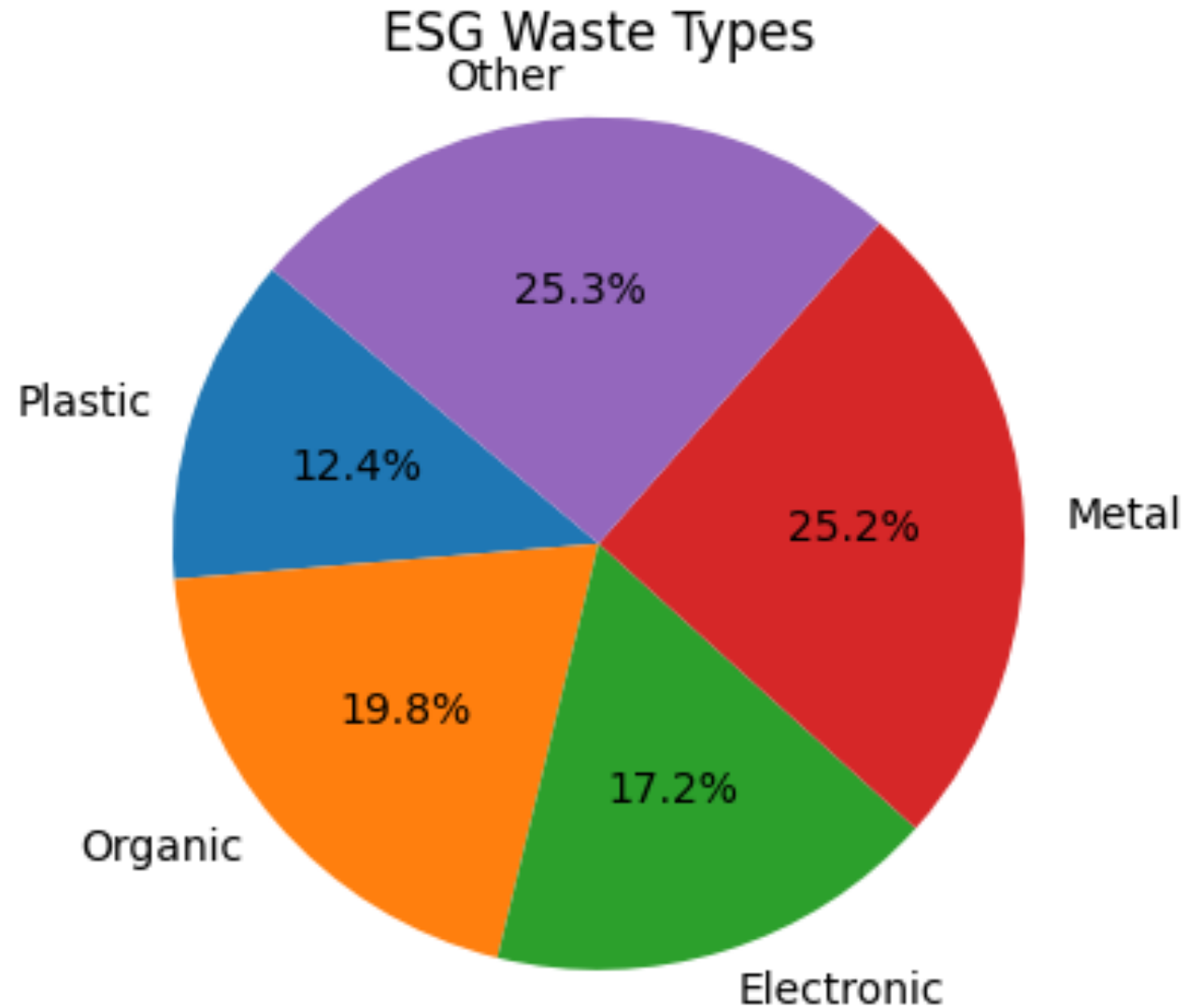
# ESG Data Analysis and Visualization



# ESG Data Analysis and Visualization

```
# Pie Chart for Waste Types
plt.figure(figsize=(8, 4))
plt.pie(df['waste_amount'], labels=df['waste_type'], autopct='%1.1f%%',
startangle=140)
plt.title('ESG Waste Types')
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a
circle.
plt.show()
plt.savefig('ESG_Waste_Type_Breakdown.jpg', format='jpg', dpi=300)
```

# ESG Data Analysis and Visualization





# Python

# Pandas



**Python**  
**matplotlib**  
*matplotlib*

# Python seaborn



seaborn

# Python plotly



# Python

# bokeh



# Python matplotlib



Fork me on GitHub

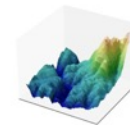
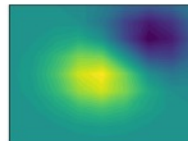
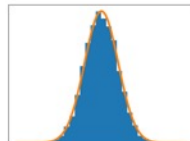
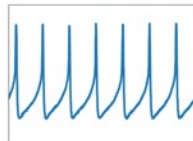
[Installation](#) [Documentation](#) [Examples](#) [Tutorials](#) [Contributing](#)

[home](#) | [contents](#) » [Matplotlib: Python plotting](#)

[modules](#) | [index](#)

## Matplotlib: Visualization with Python

Matplotlib is a comprehensive library for creating static, animated, and interactive visualizations in Python.



Matplotlib makes easy things easy and hard things possible.

### Create

- Develop **publication quality plots** with just a few lines of code
- Use **interactive figures** that can zoom, pan, update...

### Customize

- **Take full control** of line styles, font properties, axes properties...
- **Export and embed** to a number of file formats and interactive environments

### Extend

- Explore tailored functionality provided by **third party packages**
- Learn more about Matplotlib through the many **external learning resources**

#### Latest stable release

3.3.4: [docs](#) | [changelog](#)

#### Last release for Python 2

2.2.5: [docs](#) | [changelog](#)

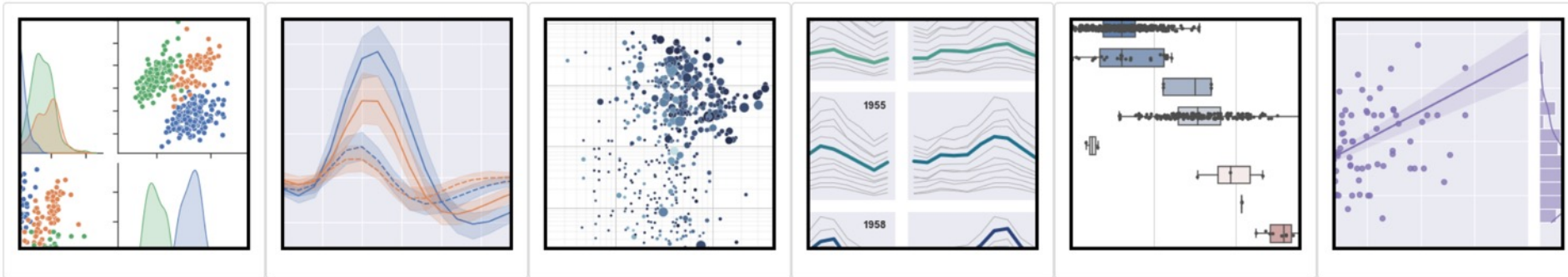
#### Development version

[docs](#)

Support Matplotlib

<https://matplotlib.org/>

# seaborn: statistical data visualization



Seaborn is a Python data visualization library based on [matplotlib](#). It provides a high-level interface for drawing attractive and informative statistical graphics.

For a brief introduction to the ideas behind the library, you can read the [introductory notes](#). Visit the [installation page](#) to see how you can download the package and get started with it. You can browse the [example gallery](#) to see what you can do with seaborn, and then check out the [tutorial](#) and [API reference](#) to find out how.

To see the code or report a bug, please visit the [GitHub repository](#). General support questions are most at home on [stackoverflow](#) or [discourse](#), which have dedicated channels for seaborn.

## Contents

- [Introduction](#)
- [Release notes](#)
- [Installing](#)
- [Example gallery](#)
- [Tutorial](#)
- [API reference](#)

## Features

- Relational: [API](#) | [Tutorial](#)
- Distribution: [API](#) | [Tutorial](#)
- Categorical: [API](#) | [Tutorial](#)
- Regression: [API](#) | [Tutorial](#)
- Multiples: [API](#) | [Tutorial](#)
- Style: [API](#) | [Tutorial](#)
- Color: [API](#) | [Tutorial](#)

# Python Plotly Graphing Library

Search...

## Quick Start

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[community.plotly.com](#)

## Examples

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[Statistical Charts](#)
[Artificial Intelligence and Machine Learning](#)
[Scientific Charts](#)
[Financial Charts](#)
[Maps](#)
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## Plotly Python Open Source Graphing Library

Plotly's Python graphing library makes interactive, publication-quality graphs. Examples of how to make line plots, scatter plots, area charts, bar charts, error bars, box plots, histograms, heatmaps, subplots, multiple-axes, polar charts, and bubble charts.

Plotly.py is [free and open source](#) and you can [view the source](#), [report issues](#) or [contribute on GitHub](#).

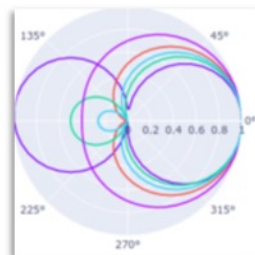
Our recommended IDE for Plotly's Python graphing library is Dash Enterprise's [Data Science Workspaces](#), which has both Jupyter notebook and Python code file support. [Find out if your company is using Dash Enterprise.](#)

[Install Dash Enterprise on Azure](#) | [Install Dash Enterprise on AWS](#)

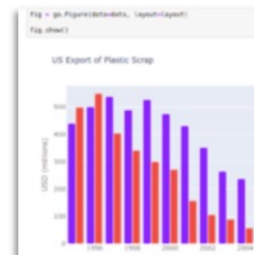
## Fundamentals

[More Fundamentals »](#)

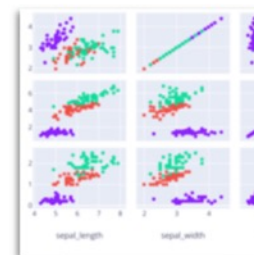

The Figure Data Structure



Creating and Updating Figures



Displaying Figures



Plotly Express



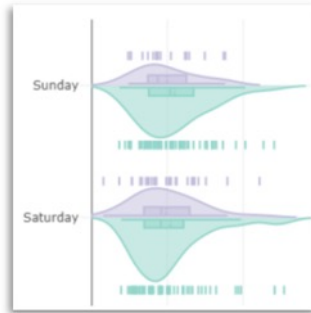
Analytical Apps with Dash



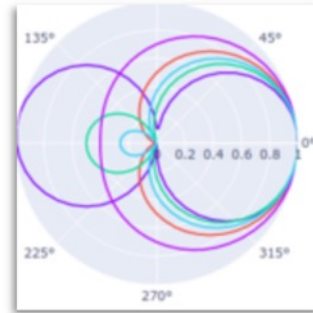
# Python Plotly Graphing Library

## Fundamentals

[More Fundamentals »](#)



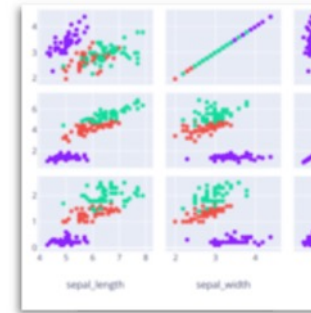
The Figure Data Structure



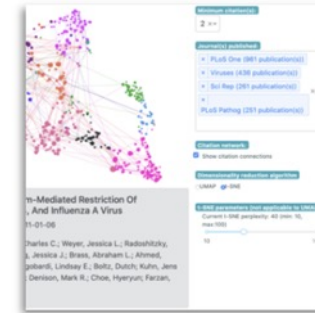
Creating and Updating Figures



Displaying Figures



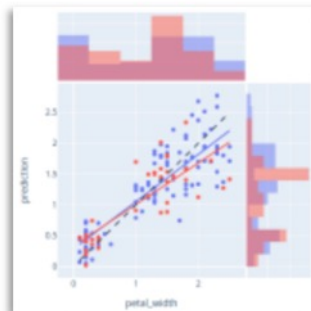
Plotly Express



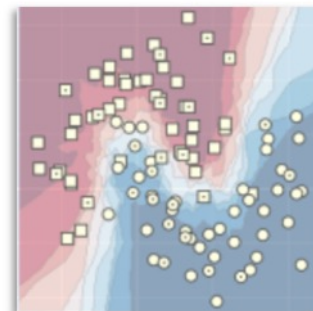
Analytical Apps with Dash

## Artificial Intelligence and Machine Learning

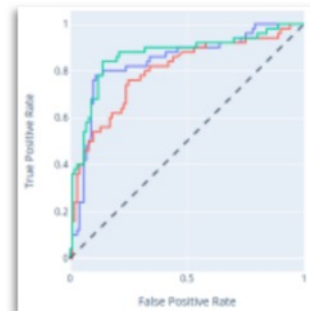
[More AI and ML »](#)



ML Regression



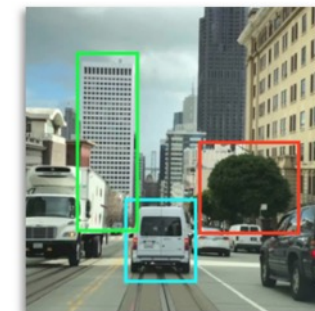
kNN Classification



ROC and PR Curves



PCA Visualization

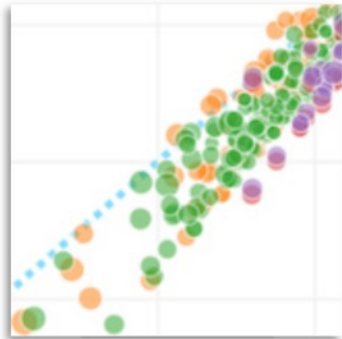


AI/ML Apps with Dash

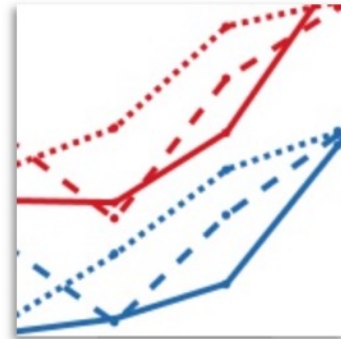
# Python Plotly Graphing Library

## Basic Charts

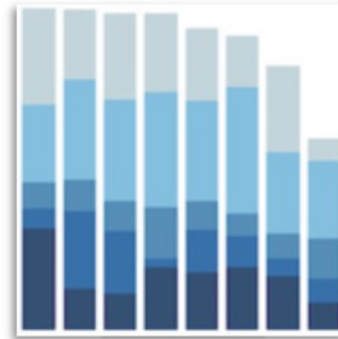
[More Basic Charts »](#)



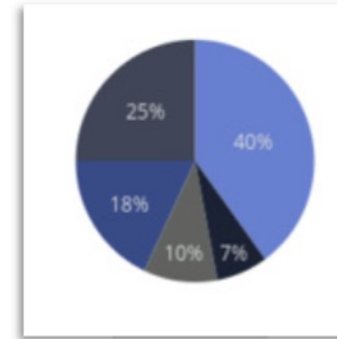
Scatter Plots



Line Charts



Bar Charts



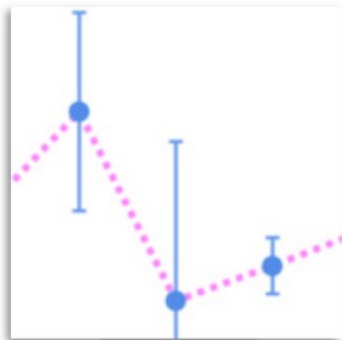
Pie Charts



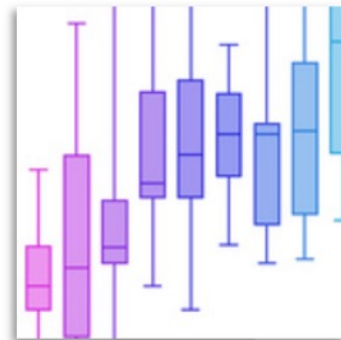
Bubble Charts

## Statistical Charts

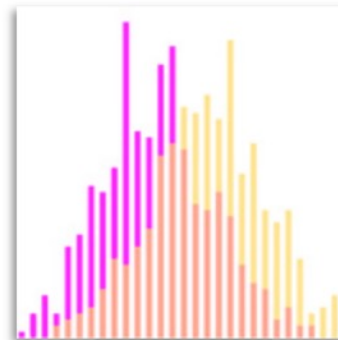
[More Statistical Charts »](#)



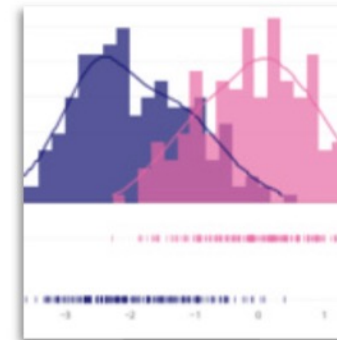
Error Bars



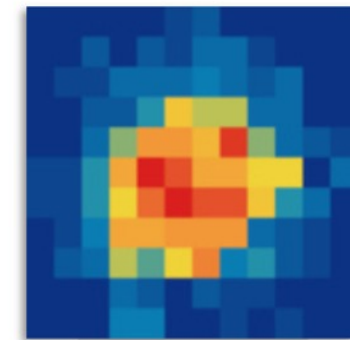
Box Plots



Histograms



Distplots

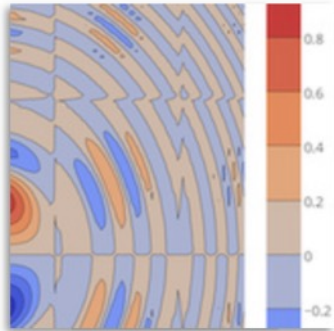


[2D Histograms](#)

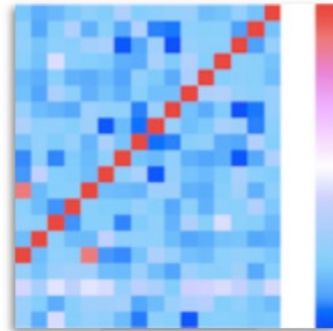
# Python Plotly Graphing Library

## Scientific Charts

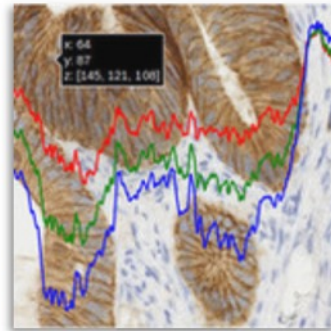
[More Scientific Charts »](#)



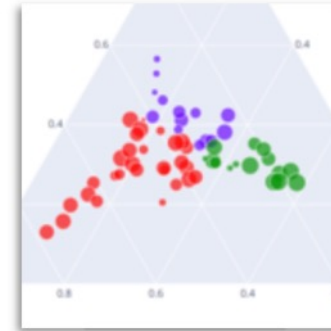
Contour Plots



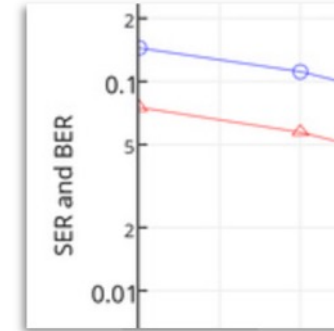
Heatmaps



Imshow



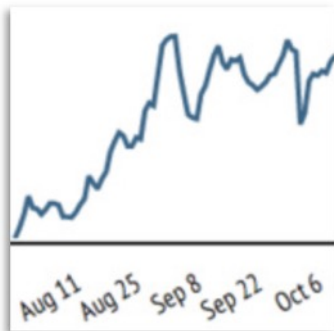
Ternary Plots



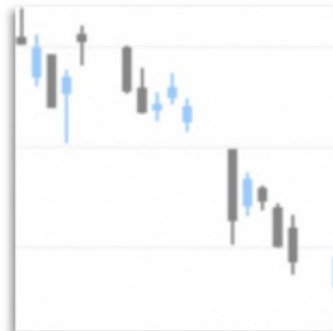
Log Plots

## Financial Charts

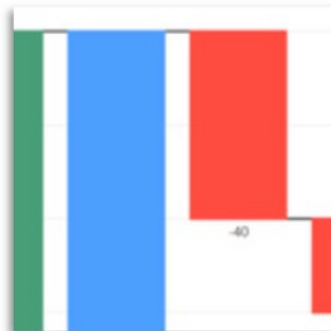
[More Financial Charts »](#)



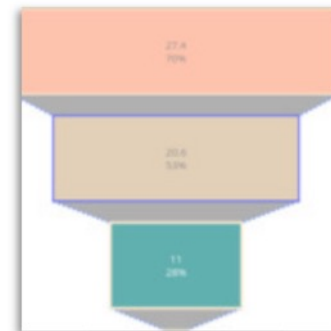
Time Series and Date  
Axes



Candlestick Charts



Waterfall Charts



Funnel Chart



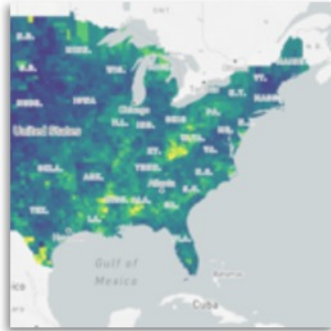
[OHLC Charts](#)



# Python Plotly Graphing Library

## Maps

[More Maps »](#)



Mapbox Choropleth Maps



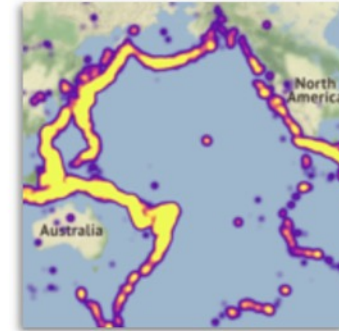
Lines on Mapbox



Filled Area on Maps



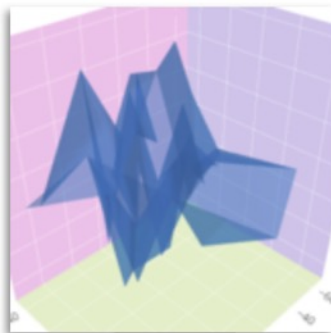
Bubble Maps



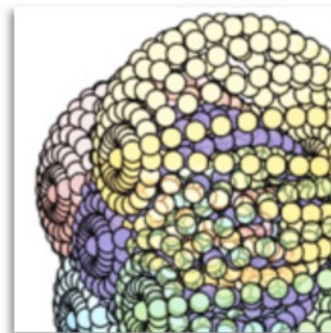
Mapbox Density Heatmap

## 3D Charts

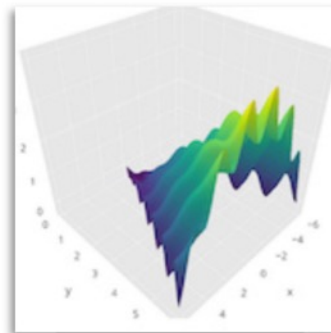
[More 3D Charts »](#)



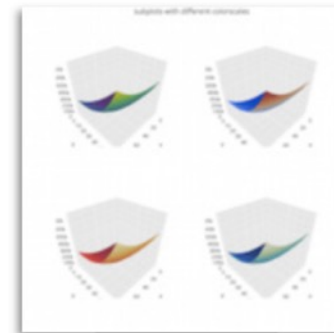
3D Axes



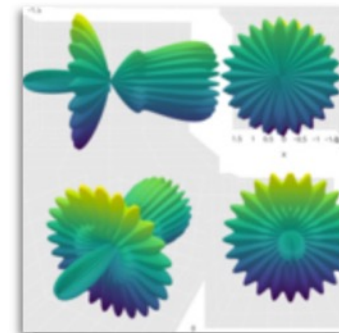
3D Scatter Plots



3D Surface Plots



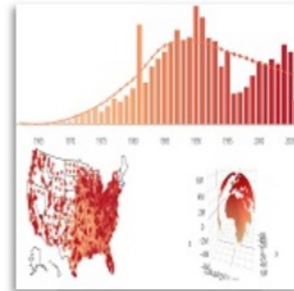
3D Subplots



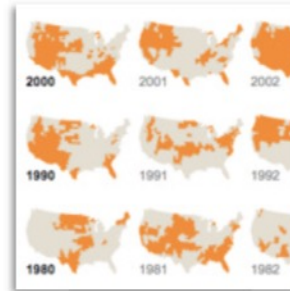
[3D Camera Controls](#)

# Python Plotly Graphing Library

## Subplots



Mixed Subplots



Map Subplots

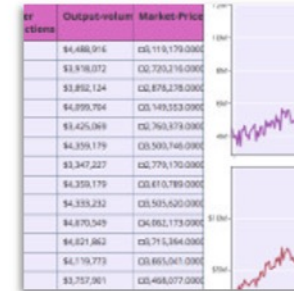
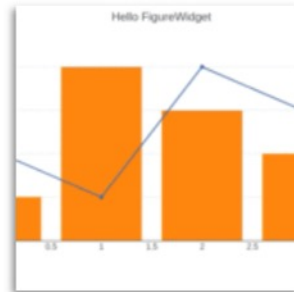


Table and Chart Subplots

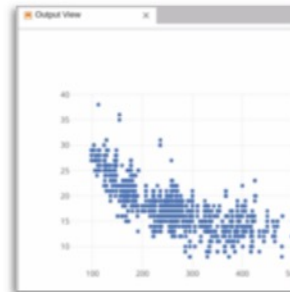


Figure Factory Subplots

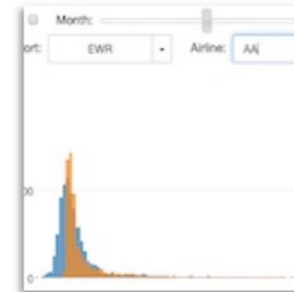
## Jupyter Widgets Interaction



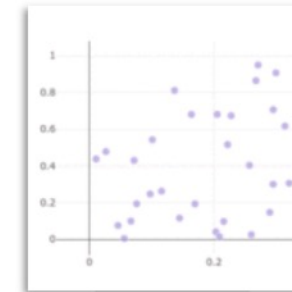
Plotly FigureWidget Overview



Jupyter Lab with FigureWidget



Interactive Data Analysis with FigureWidget ipywidgets



Click Events

# bokeh Python Bokeh



2.3.0 ▾

First steps

User guide

Gallery

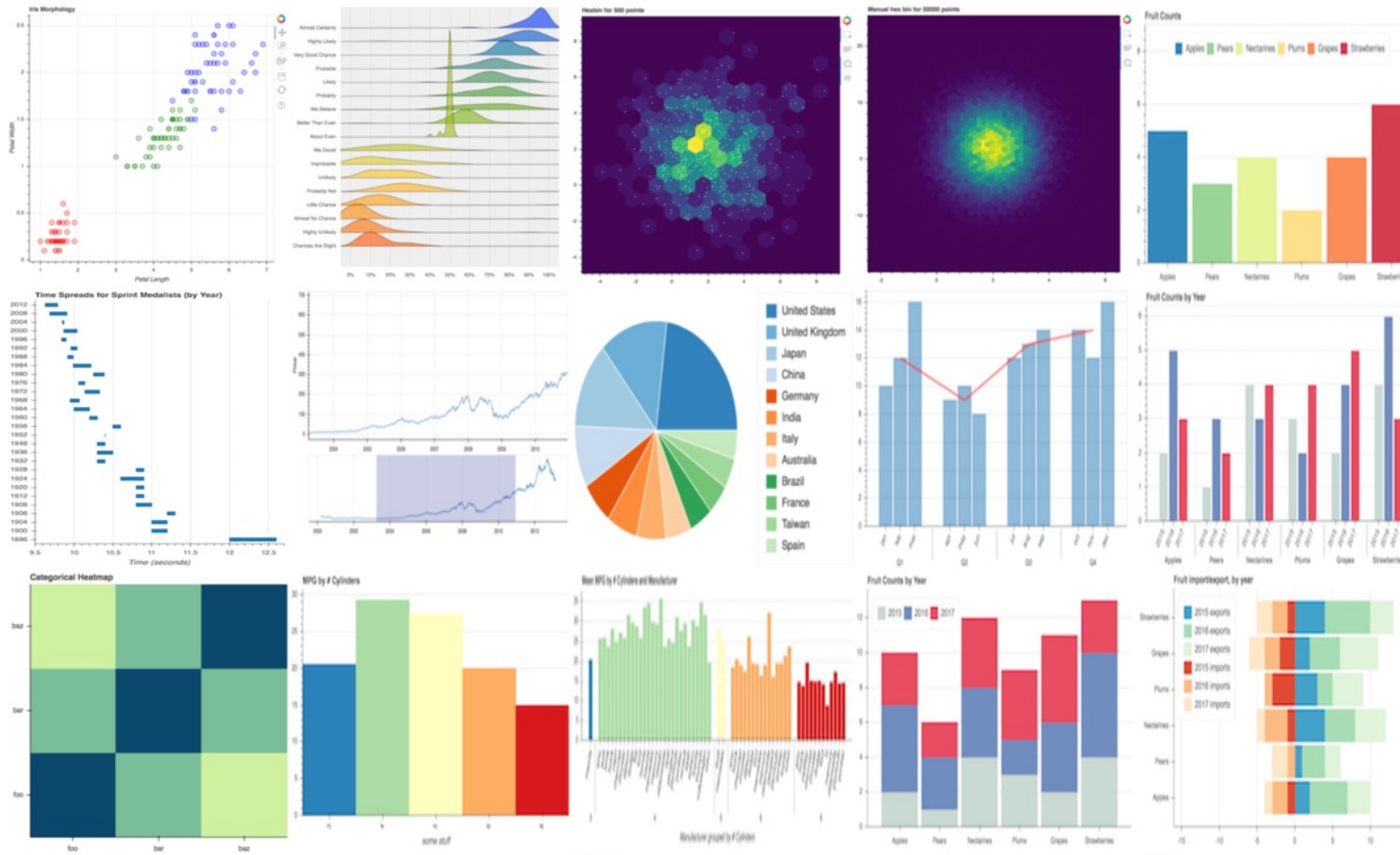
Reference

Developers

Releases

Tutorial ↗

Community ↗



# Python in Google Colab (Python101)

<https://colab.research.google.com/drive/1FEG6DnGvwfUbeo4zJ1zTunjMqf2RkCrT>

python101.ipynb ☆

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  - Investment Portfolio Optimisation with Python
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+ Code + Text

```
[2] 1 import seaborn as sns
2 sns.set(style="ticks", color_codes=True)
3 iris = sns.load_dataset("iris")
4 g = sns.pairplot(iris, hue="species")
```

<https://tinyurl.com/aintpupython101>

# Summary

- **ESG Data Gathering**
- **ESG Data Analysis**
- **ESG Data Visualization**



# References

- Yi Zou, Mengying Shi, Zhongjie Chen, Zhu Deng, ZongXiong Lei, Zihan Zeng, Shiming Yang, Hongxiang Tong, Lei Xiao, and Wenwen Zhou. (2025) "ESGReveal: An LLM-based approach for extracting structured data from ESG reports." Journal of Cleaner Production 489 (2025): 144572.
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- Min-Yuh Day (2025), Python 101, <https://tinyurl.com/aintpupython101>