

Phase 2

Group Members: Anusha Bishayee, Katheryn Ding, Phoebe Wang, Kaiwen Zhou

Research Question:

How do ESG performance and stock performance correlate across industries? What associations can we find between industry, stock performance, and ESG rating?

Data Collection and Cleaning

```
In [206... import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
from sklearn.linear_model import LinearRegression
esg = pd.read_csv("esg_data.csv")
print(esg.iloc[0:15,:])
```

	ticker	name	currency	\
0	dis	Walt Disney Co	USD	
1	gm	General Motors Co	USD	
2	gww	WW Grainger Inc	USD	
3	mhk	Mohawk Industries Inc	USD	
4	lyv	Live Nation Entertainment Inc	USD	
5	lvs	Las Vegas Sands Corp	USD	
6	clx	Clorox Co	USD	
7	aacg	ATA Creativity Global	CNY	
8	aal	American Airlines Group Inc	USD	
9	aame	Atlantic American Corp	USD	
10	aaoi	Applied Optoelectronics Inc	USD	

11	aaon	Aaon Inc	USD
12	aapl	Apple Inc	USD
13	aatc	Autoscope Technologies Corp	USD
14	aaww	Atlas Air Worldwide Holdings Inc	USD

	exchange	industry \
0	NEW YORK STOCK EXCHANGE, INC.	Media
1	NEW YORK STOCK EXCHANGE, INC.	Automobiles
2	NEW YORK STOCK EXCHANGE, INC.	Trading Companies and Distributors
3	NEW YORK STOCK EXCHANGE, INC.	Consumer products
4	NEW YORK STOCK EXCHANGE, INC.	Media
5	NEW YORK STOCK EXCHANGE, INC.	Hotels Restaurants and Leisure
6	NEW YORK STOCK EXCHANGE, INC.	Consumer products
7	NASDAQ NMS - GLOBAL MARKET	Diversified Consumer Services
8	NASDAQ NMS - GLOBAL MARKET	Airlines
9	NASDAQ NMS - GLOBAL MARKET	Insurance
10	NASDAQ NMS - GLOBAL MARKET	Communications
11	NASDAQ NMS - GLOBAL MARKET	Building
12	NASDAQ NMS - GLOBAL MARKET	Technology
13	NASDAQ NMS - GLOBAL MARKET	Electrical Equipment
14	NASDAQ NMS - GLOBAL MARKET	Logistics and Transportation

	logo \
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2	https://static.finnhub.io/logo/f153dcda-80eb-1...
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13	NaN
14	https://static.finnhub.io/logo/8b74bf8a-80df-1...

weburldata environment_grade \

0	https://thewaltdisneycompany.com/	A
1	https://www.gm.com/	A
2	https://www.grainger.com/	B
3	https://mohawkind.com/	A
4	https://www.livenationentertainment.com/	BBB
5	https://www.sands.com/	A
6	https://www.thecloroxcompany.com/	A
7	http://www.ata.net.cn	B
8	https://americanairlines.gcs-web.com/	B
9	http://www.atlam.com/	B
10	http://ao-inc.com/	BB
11	https://www.aaon.com/	A
12	https://www.apple.com/	BB
13	NaN	BBB
14	https://www.atlasairworldwide.com/	A

	environment_level	social_grade	...	governance_grade	governance_level	\
0	High	BB	...	BB	Medium	
1	High	BB	...	B	Medium	
2	Medium	BB	...	B	Medium	
3	High	B	...	BB	Medium	
4	High	BB	...	B	Medium	
5	High	BB	...	BB	Medium	
6	High	BB	...	BB	Medium	
7	Medium	B	...	B	Medium	
8	Medium	B	...	B	Medium	
9	Medium	B	...	BB	Medium	
10	Medium	B	...	B	Medium	
11	High	BB	...	B	Medium	
12	Medium	B	...	B	Medium	
13	High	B	...	B	Medium	
14	High	B	...	B	Medium	

	environment_score	social_score	governance_score	total_score	\
0	510	316	321	1147	
1	510	303	255	1068	
2	255	385	240	880	
3	570	298	303	1171	
4	492	310	250	1052	
5	547	318	313	1178	

6	560	350	345	1255
7	203	200	205	608
8	270	211	265	746
9	220	221	300	741
10	311	256	218	785
11	500	300	223	1023
12	355	281	255	891
13	410	286	240	936
14	520	243	237	1000

	last_processing_date	total_grade	total_level	cik
0	19-04-2022	BBB	High	1744489
1	17-04-2022	BBB	High	1467858
2	19-04-2022	BB	Medium	277135
3	18-04-2022	BBB	High	851968
4	18-04-2022	BBB	High	1335258
5	18-04-2022	BBB	High	1300514
6	16-04-2022	A	High	21076
7	01-07-2022	B	Medium	1420529
8	16-04-2022	B	Medium	6201
9	16-04-2022	B	Medium	8177
10	16-04-2022	BB	Medium	1158114
11	16-04-2022	BBB	High	824142
12	16-04-2022	BB	Medium	320193
13	11-06-2022	BBB	High	943034
14	16-04-2022	BBB	High	1135185

[15 rows x 21 columns]

```
In [207... #cleaning to only have certain companies that represent a variety of industries

companies = ["Walt Disney Co", "American Airlines Group Inc", "Apple Inc", "eBay Inc", "Goldman Sachs Gro
            "Meta Platforms Inc", "Starbucks Corp", "PayPal Holdings Inc", "United Airlines Holdings Inc",
            "Bath & Body Works Inc", "Abbvie Inc", "Alexandria Real Estate Equities Inc",
            "Becton Dickinson and Co", "Brown & Brown Inc", "Duke Energy Corp", "T-Mobile US Inc",
            "Marathon Oil Corp", "Chipotle Mexican Grill Inc", "Target Corp",
            "General Motors Co", "Salesforce Inc", "Tesla Inc", "Bank of America Corp"]

relevant_esg = esg[esg["name"].isin(companies)]
print(relevant_esg)
```

	ticker	name	currency \
0	dis	Walt Disney Co	USD
1	gm	General Motors Co	USD
8	aal	American Airlines Group Inc	USD
12	aapl	Apple Inc	USD
163	ebay	eBay Inc	USD
170	meta	Meta Platforms Inc	USD
196	sbux	Starbucks Corp	USD
201	pypl	PayPal Holdings Inc	USD
212	tsla	Tesla Inc	USD
215	tmus	T-Mobile US Inc	USD
216	ual	United Airlines Holdings Inc	USD
231	abbv	Abbvie Inc	USD
256	are	Alexandria Real Estate Equities Inc	USD
261	bac	Bank of America Corp	USD
263	bbwi	Bath & Body Works Inc	USD
264	bdx	Becton Dickinson and Co	USD
271	bro	Brown & Brown Inc	USD
289	cmg	Chipotle Mexican Grill Inc	USD
295	duk	Duke Energy Corp	USD
306	gs	Goldman Sachs Group Inc	USD
333	mro	Marathon Oil Corp	USD
399	tgt	Target Corp	USD
445	crm	Salesforce Inc	USD

	exchange	industry \
0	NEW YORK STOCK EXCHANGE, INC.	Media
1	NEW YORK STOCK EXCHANGE, INC.	Automobiles
8	NASDAQ NMS - GLOBAL MARKET	Airlines
12	NASDAQ NMS - GLOBAL MARKET	Technology
163	NASDAQ NMS - GLOBAL MARKET	Retail
170	NASDAQ NMS - GLOBAL MARKET	Media
196	NASDAQ NMS - GLOBAL MARKET	Hotels Restaurants and Leisure
201	NASDAQ NMS - GLOBAL MARKET	Technology
212	NASDAQ NMS - GLOBAL MARKET	Automobiles
215	NASDAQ NMS - GLOBAL MARKET	Telecommunication
216	NASDAQ NMS - GLOBAL MARKET	Airlines
231	NEW YORK STOCK EXCHANGE, INC.	Biotechnology
256	NEW YORK STOCK EXCHANGE, INC.	Real Estate
261	NEW YORK STOCK EXCHANGE, INC.	Banking

263	NEW YORK STOCK EXCHANGE, INC.	Retail
264	NEW YORK STOCK EXCHANGE, INC.	Health Care
271	NEW YORK STOCK EXCHANGE, INC.	Insurance
289	NEW YORK STOCK EXCHANGE, INC.	Hotels Restaurants and Leisure
295	NEW YORK STOCK EXCHANGE, INC.	Utilities
306	NEW YORK STOCK EXCHANGE, INC.	Financial Services
333	NEW YORK STOCK EXCHANGE, INC.	Energy
399	NEW YORK STOCK EXCHANGE, INC.	Retail
445	NEW YORK STOCK EXCHANGE, INC.	Technology

	logo \
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1	https://static.finnhub.io/logo/9253db78-80c9-1...
8	https://static2.finnhub.io/file/publicdatany/f...
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216	https://static.finnhub.io/logo/b3f34b67dcba172...
231	https://static.finnhub.io/logo/8806d72c-80cd-1...
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264	https://static.finnhub.io/logo/82e1f20c-80eb-1...
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399	https://static.finnhub.io/logo/83bbf858-80ec-1...
445	https://static.finnhub.io/logo/5a800a68f67c85e...

	weblink environment_grade \
0	https://thewaltdisneycompany.com/ A
1	https://www.gm.com/ A
8	https://americanairlines.gcs-web.com/ B
12	https://www.apple.com/ BB

163	https://www.ebayinc.com/	A
170	https://www.facebook.com	B
196	https://www.starbucks.com/	BBB
201	https://www.paypal.com/	BB
212	https://www.tesla.com/	A
215	https://www.t-mobile.com/	BB
216	https://ir.united.com/	B
231	https://www.abbvie.com/	A
256	https://www.are.com/	AA
261	https://www.bankofamerica.com	B
263	https://www.bbwin.com/	A
264	https://www.bd.com/	AA
271	https://www.bbinsurance.com/	BBB
289	https://www.chipotle.com/	B
295	https://www.duke-energy.com/	AA
306	https://www.goldmansachs.com/	A
333	https://www.marathonoil.com/	A
399	https://corporate.target.com/	A
445	https://www.salesforce.com/	B

	environment_level	social_grade	...	governance_grade	governance_level	\
0	High	BB	...	BB	Medium	
1	High	BB	...	B	Medium	
8	Medium	B	...	B	Medium	
12	Medium	B	...	B	Medium	
163	High	BB	...	BB	Medium	
170	Medium	B	...	B	Medium	
196	High	BB	...	B	Medium	
201	Medium	B	...	B	Medium	
212	High	CCC	...	B	Medium	
215	Medium	BB	...	B	Medium	
216	Medium	B	...	B	Medium	
231	High	BB	...	BB	Medium	
256	Excellent	BB	...	BB	Medium	
261	Medium	BB	...	B	Medium	
263	High	BB	...	BB	Medium	
264	Excellent	BB	...	BB	Medium	
271	High	BB	...	BB	Medium	
289	Medium	B	...	B	Medium	
295	Excellent	BB	...	BB	Medium	

306	High	BB	...	BB	Medium
333	High	BB	...	BB	Medium
399	High	BB	...	BB	Medium
445	Medium	B	...	B	Medium

	environment_score	social_score	governance_score	total_score	\
0	510	316	321	1147	
1	510	303	255	1068	
8	270	211	265	746	
12	355	281	255	891	
163	500	302	305	1107	
170	205	232	215	652	
196	475	385	295	1155	
201	370	244	285	899	
212	555	160	278	993	
215	342	331	230	903	
216	205	203	221	629	
231	505	317	300	1122	
256	600	337	345	1282	
261	220	394	265	879	
263	525	303	300	1128	
264	607	384	315	1306	
271	405	320	300	1025	
289	210	241	213	664	
295	604	362	328	1294	
306	525	305	305	1135	
333	530	313	305	1148	
399	550	342	313	1205	
445	215	241	200	656	

	last_processing_date	total_grade	total_level	cik
0	19-04-2022	BBB	High	1744489
1	17-04-2022	BBB	High	1467858
8	16-04-2022	B	Medium	6201
12	16-04-2022	BB	Medium	320193
163	17-04-2022	BBB	High	1065088
170	22-04-2022	B	Medium	1326801
196	19-04-2022	BBB	High	829224
201	18-04-2022	BB	Medium	1633917
212	19-04-2022	BBB	High	1318605

215	19-04-2022	BBB	High	1283699
216	19-04-2022	B	Medium	100517
231	16-04-2022	BBB	High	1551152
256	16-04-2022	A	High	1035443
261	16-04-2022	BB	Medium	70858
263	11-06-2022	BBB	High	701985
264	16-04-2022	A	High	10795
271	16-04-2022	BBB	High	79282
289	16-04-2022	B	Medium	1058090
295	17-04-2022	A	High	1326160
306	17-04-2022	BBB	High	886982
333	21-04-2022	BBB	High	101778
399	19-04-2022	A	High	27419
445	19-04-2022	B	Medium	1108524

[23 rows x 21 columns]

In [208... `import yfinance as yf`

In [209... `#use yfinance to pull stock information of selected stocks.`

```

relevant_esg.loc[:, 'ticker'] = relevant_esg['ticker'].astype(str)
relevant_esg.loc[:, 'name'] = relevant_esg['name'].astype(str)
tickers = relevant_esg['ticker'].tolist()
#Add Start Price, End Price, and Rate of Change (%) of each company to the dataset relevant.esg
relevant_esg.loc[:, 'Start Price'] = None
relevant_esg.loc[:, 'End Price'] = None
relevant_esg.loc[:, 'Rate of Change (%)'] = None

# Loop through each row of the DataFrame to get stock information for each company
for index, row in relevant_esg.iterrows():
    ticker = row['ticker']

    # Download stock data for 2023
    data = yf.download(ticker, start='2021-04-01', end='2022-04-01')

    # Ensure data exists for the given period
    if not data.empty:
        start_price = data['Adj Close'].iloc[0]
        end_price = data['Adj Close'].iloc[-1]
        rate_of_change = ((end_price - start_price) / start_price) * 100

```

```
print(relevant_esg)
```

	ticker	name	currency
0	dis	Walt Disney Co	USD
1	gm	General Motors Co	USD
8	aal	American Airlines Group Inc	USD
12	aapl	Apple Inc	USD
163	ebay	eBay Inc	USD
170	meta	Meta Platforms Inc	USD
196	sbux	Starbucks Corp	USD
201	pypl	PayPal Holdings Inc	USD

212	tsla	Tesla Inc	USD
215	tmus	T-Mobile US Inc	USD
216	ual	United Airlines Holdings Inc	USD
231	abbv	Abbvie Inc	USD
256	are	Alexandria Real Estate Equities Inc	USD
261	bac	Bank of America Corp	USD
263	bbwi	Bath & Body Works Inc	USD
264	bdx	Becton Dickinson and Co	USD
271	bro	Brown & Brown Inc	USD
289	cmg	Chipotle Mexican Grill Inc	USD
295	duk	Duke Energy Corp	USD
306	gs	Goldman Sachs Group Inc	USD
333	mro	Marathon Oil Corp	USD
399	tgt	Target Corp	USD
445	crm	Salesforce Inc	USD

	exchange	industry \
0	NEW YORK STOCK EXCHANGE, INC.	Media
1	NEW YORK STOCK EXCHANGE, INC.	Automobiles
8	NASDAQ NMS - GLOBAL MARKET	Airlines
12	NASDAQ NMS - GLOBAL MARKET	Technology
163	NASDAQ NMS - GLOBAL MARKET	Retail
170	NASDAQ NMS - GLOBAL MARKET	Media
196	NASDAQ NMS - GLOBAL MARKET	Hotels Restaurants and Leisure
201	NASDAQ NMS - GLOBAL MARKET	Technology
212	NASDAQ NMS - GLOBAL MARKET	Automobiles
215	NASDAQ NMS - GLOBAL MARKET	Telecommunication
216	NASDAQ NMS - GLOBAL MARKET	Airlines
231	NEW YORK STOCK EXCHANGE, INC.	Biotechnology
256	NEW YORK STOCK EXCHANGE, INC.	Real Estate
261	NEW YORK STOCK EXCHANGE, INC.	Banking
263	NEW YORK STOCK EXCHANGE, INC.	Retail
264	NEW YORK STOCK EXCHANGE, INC.	Health Care
271	NEW YORK STOCK EXCHANGE, INC.	Insurance
289	NEW YORK STOCK EXCHANGE, INC.	Hotels Restaurants and Leisure
295	NEW YORK STOCK EXCHANGE, INC.	Utilities
306	NEW YORK STOCK EXCHANGE, INC.	Financial Services
333	NEW YORK STOCK EXCHANGE, INC.	Energy
399	NEW YORK STOCK EXCHANGE, INC.	Retail
445	NEW YORK STOCK EXCHANGE, INC.	Technology

	logo \
0	https://static.finnhub.io/logo/ef50b4a2b263c84...
1	https://static.finnhub.io/logo/9253db78-80c9-1...
8	https://static2.finnhub.io/file/publicdatany/f...
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399	https://static.finnhub.io/logo/83bbf858-80ec-1...
445	https://static.finnhub.io/logo/5a800a68f67c85e...

	weburl	environment_grade \
0	https://thewaltdisneycompany.com/	A
1	https://www.gm.com/	A
8	https://americanairlines.gcs-web.com/	B
12	https://www.apple.com/	BB
163	https://www.ebayinc.com/	A
170	https://www.facebook.com	B
196	https://www.starbucks.com/	BBB
201	https://www.paypal.com/	BB
212	https://www.tesla.com/	A
215	https://www.t-mobile.com/	BB
216	https://ir.united.com/	B
231	https://www.abbvie.com/	A
256	https://www.are.com/	AA

261	https://www.bankofamerica.com	B
263	https://www.bbwin.com/	A
264	https://www.bd.com/	AA
271	https://www.bbinsurance.com/	BBB
289	https://www.chipotle.com/	B
295	https://www.duke-energy.com/	AA
306	https://www.goldmansachs.com/	A
333	https://www.marathonoil.com/	A
399	https://corporate.target.com/	A
445	https://www.salesforce.com/	B

	environment_level	social_grade	...	social_score	governance_score	\
0	High	BB	...	316	321	
1	High	BB	...	303	255	
8	Medium	B	...	211	265	
12	Medium	B	...	281	255	
163	High	BB	...	302	305	
170	Medium	B	...	232	215	
196	High	BB	...	385	295	
201	Medium	B	...	244	285	
212	High	CCC	...	160	278	
215	Medium	BB	...	331	230	
216	Medium	B	...	203	221	
231	High	BB	...	317	300	
256	Excellent	BB	...	337	345	
261	Medium	BB	...	394	265	
263	High	BB	...	303	300	
264	Excellent	BB	...	384	315	
271	High	BB	...	320	300	
289	Medium	B	...	241	213	
295	Excellent	BB	...	362	328	
306	High	BB	...	305	305	
333	High	BB	...	313	305	
399	High	BB	...	342	313	
445	Medium	B	...	241	200	

	total_score	last_processing_date	total_grade	total_level	cik	\
0	1147	19-04-2022	BBB	High	1744489	
1	1068	17-04-2022	BBB	High	1467858	
8	746	16-04-2022	B	Medium	6201	

12	891	16-04-2022	BB	Medium	320193
163	1107	17-04-2022	BBB	High	1065088
170	652	22-04-2022	B	Medium	1326801
196	1155	19-04-2022	BBB	High	829224
201	899	18-04-2022	BB	Medium	1633917
212	993	19-04-2022	BBB	High	1318605
215	903	19-04-2022	BBB	High	1283699
216	629	19-04-2022	B	Medium	100517
231	1122	16-04-2022	BBB	High	1551152
256	1282	16-04-2022	A	High	1035443
261	879	16-04-2022	BB	Medium	70858
263	1128	11-06-2022	BBB	High	701985
264	1306	16-04-2022	A	High	10795
271	1025	16-04-2022	BBB	High	79282
289	664	16-04-2022	B	Medium	1058090
295	1294	17-04-2022	A	High	1326160
306	1135	17-04-2022	BBB	High	886982
333	1148	21-04-2022	BBB	High	101778
399	1205	19-04-2022	A	High	27419
445	656	19-04-2022	B	Medium	1108524

	Start Price	End Price	Rate of Change (%)
0	187.491135	136.086594	-27.417052
1	56.471729	42.734833	-24.325263
8	23.860001	18.25	-23.512156
12	120.588425	172.203995	42.803088
163	59.152027	54.319157	-8.170253
170	297.763458	221.692505	-25.547444
196	101.377022	85.777168	-15.387958
201	247.539993	115.650002	-53.280276
212	220.583328	359.200012	62.840961
215	125.69912	126.388428	0.548379
216	57.82	46.360001	-19.82013
231	93.50222	146.184769	56.343634
256	149.982285	181.9478	21.312861
261	36.186047	38.49548	6.382109
263	46.90102	45.340763	-3.326702
264	223.961151	249.956436	11.607051
271	45.428921	71.043533	56.383934
289	28.769199	31.6406	9.980816

295	83.651123	100.818703	20.522832
306	299.524689	307.67276	2.720334
333	11.18561	24.169741	116.078882
399	184.384384	197.776855	7.263344
445	217.753311	211.381607	-2.926111

[23 rows x 24 columns]

```
In [210... #Add Start Price, End Price, and Rate of Change (%) of each company to the dataset esg
esg.loc[:, 'ticker'] = esg['ticker'].astype(str)
esg.loc[:, 'name'] = esg['name'].astype(str)
tickers = esg['ticker'].tolist()

# Add columns for Start Price, End Price, and Rate of Change (%)
esg['Start Price'] = None
esg['End Price'] = None
esg['Rate of Change (%)'] = None

# Loop through each row of the DataFrame to get stock information for each company
for index, row in esg.iterrows():
    ticker = row['ticker']

    # Skip rows where the ticker is not valid
    if ticker == 'nan' or ticker.strip() == '':
        continue

    # Download stock data for the given ticker
    data = yf.download(ticker, start='2021-04-01', end='2022-04-01')

    # Ensure data exists for the given period
    if not data.empty:
        start_price = data['Adj Close'].iloc[0]
        end_price = data['Adj Close'].iloc[-1]
        rate_of_change = ((end_price - start_price) / start_price) * 100

    # Add the stock data to the relevant columns in the DataFrame
    esg.loc[index, 'Start Price'] = start_price
    esg.loc[index, 'End Price'] = end_price
    esg.loc[index, 'Rate of Change (%)'] = rate_of_change
```

```
#Remove companies with missing value on stock information:
esg.dropna(subset=['Start Price', 'End Price', 'Rate of Change (%)'], inplace=True)

#show the first 15 rows of cleaned esg
print(esg.iloc[0:15,:])
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1 Failed download:

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['ABMD']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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1 Failed download:

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['ACAC']: YFPricesMissingError('$%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01) (Yahoo error = "Data doesn\'t exist for startDate = 1617249600, endDate = 1648785600")')
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['ACEV']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['ADES']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['ADMP']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['ACER']: YFPricesMissingError('$%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01)')
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['AHPI']: YFPricesMissingError('%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01)')
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['AIKI']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['AKU']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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1 Failed download:

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['AKUS']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['ALF']: YFPricesMissingError('$%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01) (Yahoo error = "Data doesn\'t exist for startDate = 1617249600, endDate = 1648785600")')
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[ 'CTXS' ]: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['NLOK']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['ABC']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['SIVB']: YFPricesMissingError('%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01)')
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['CDAY']: YFTzMissingError('%$ticker%: possibly delisted; no timezone found')
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['DRE']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['NLSN']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['PKI']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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['PXD']: YFTzMissingError('$%ticker%: possibly delisted; no timezone found')
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[ 'FISV' ]: YFPricesMissingError('$%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01) (Yahoo error = "Data doesn\'t exist for startDate = 1617249600, endDate = 1648785600")')
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['ZEN']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['WWE']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['XL']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['Y']: %ticker%FTzMissingError('%ticker%: possibly delisted; no timezone found')
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1 Failed download:

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['SIEN']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['MTCR']: YFPricesMissingError('%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01)')
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['HAAC']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['CLXT']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['EVK']: YFPricesMissingError('%ticker%: possibly delisted; no price data found (1d 2021-04-01 -> 2022-04-01)')
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['CLVS']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['HAPP']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['HCAR']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['HCDI']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['HCCI']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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['POW']: YFTzMissingError('%ticker%: possibly delisted; no timezone found')
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1 Failed download:

[illegible]

	ticker	name	currency
0	dis	Walt Disney Co	USD
1	gm	General Motors Co	USD
2	gww	WW Grainger Inc	USD
3	mhk	Mohawk Industries Inc	USD
4	lyv	Live Nation Entertainment Inc	USD
5	lvs	Las Vegas Sands Corp	USD
6	clx	Clorox Co	USD

7	aacg	ATA Creativity Global	CNY
8	aal	American Airlines Group Inc	USD
9	aame	Atlantic American Corp	USD
10	aaoi	Applied Optoelectronics Inc	USD
11	aaon	Aaon Inc	USD
12	aapl	Apple Inc	USD
13	aatc	Autoscope Technologies Corp	USD
15	aaci	Armada Acquisition Corp I	USD

	exchange	industry \
0	NEW YORK STOCK EXCHANGE, INC.	Media
1	NEW YORK STOCK EXCHANGE, INC.	Automobiles
2	NEW YORK STOCK EXCHANGE, INC.	Trading Companies and Distributors
3	NEW YORK STOCK EXCHANGE, INC.	Consumer products
4	NEW YORK STOCK EXCHANGE, INC.	Media
5	NEW YORK STOCK EXCHANGE, INC.	Hotels Restaurants and Leisure
6	NEW YORK STOCK EXCHANGE, INC.	Consumer products
7	NASDAQ NMS - GLOBAL MARKET	Diversified Consumer Services
8	NASDAQ NMS - GLOBAL MARKET	Airlines
9	NASDAQ NMS - GLOBAL MARKET	Insurance
10	NASDAQ NMS - GLOBAL MARKET	Communications
11	NASDAQ NMS - GLOBAL MARKET	Building
12	NASDAQ NMS - GLOBAL MARKET	Technology
13	NASDAQ NMS - GLOBAL MARKET	Electrical Equipment
15	NASDAQ NMS - GLOBAL MARKET	NaN

	logo \
0	https://static.finnhub.io/logo/ef50b4a2b263c84...
1	https://static.finnhub.io/logo/9253db78-80c9-1...
2	https://static.finnhub.io/logo/f153dcda-80eb-1...
3	https://static.finnhub.io/logo/26868a62-80ec-1...
4	https://static.finnhub.io/logo/1cd144d2-80ec-1...
5	https://static.finnhub.io/logo/7256a9be-8279-1...
6	https://static.finnhub.io/logo/a67c34fe-80eb-1...
7	https://static2.finnhub.io/file/publicdatany/f...
8	https://static2.finnhub.io/file/publicdatany/f...
9	https://static.finnhub.io/logo/6b536b3e-826a-1...
10	https://static.finnhub.io/logo/7afdb95a-80ce-1...
11	https://static.finnhub.io/logo/8819421c-80df-1...
12	https://static.finnhub.io/logo/87cb30d8-80df-1...

13	NaN
15	NaN

	weburl	environment_grade	\
0	https://thewaltdisneycompany.com/	A	
1	https://www.gm.com/	A	
2	https://www.grainger.com/	B	
3	https://mohawkind.com/	A	
4	https://www.livenationentertainment.com/	BBB	
5	https://www.sands.com/	A	
6	https://www.thecloroxcompany.com/	A	
7	http://www.ata.net.cn	B	
8	https://americanairlines.gcs-web.com/	B	
9	http://www.atlam.com/	B	
10	http://ao-inc.com/	BB	
11	https://www.aaon.com/	A	
12	https://www.apple.com/	BB	
13	NaN	BBB	
15	NaN	B	

	environment_level	social_grade	...	social_score	governance_score	\
0	High	BB	...	316	321	
1	High	BB	...	303	255	
2	Medium	BB	...	385	240	
3	High	B	...	298	303	
4	High	BB	...	310	250	
5	High	BB	...	318	313	
6	High	BB	...	350	345	
7	Medium	B	...	200	205	
8	Medium	B	...	211	265	
9	Medium	B	...	221	300	
10	Medium	B	...	256	218	
11	High	BB	...	300	223	
12	Medium	B	...	281	255	
13	High	B	...	286	240	
15	Medium	B	...	200	200	

	total_score	last_processing_date	total_grade	total_level	cik	\
0	1147	19-04-2022	BBB	High	1744489	
1	1068	17-04-2022	BBB	High	1467858	

2	880	19-04-2022	BB	Medium	277135
3	1171	18-04-2022	BBB	High	851968
4	1052	18-04-2022	BBB	High	1335258
5	1178	18-04-2022	BBB	High	1300514
6	1255	16-04-2022	A	High	21076
7	608	01-07-2022	B	Medium	1420529
8	746	16-04-2022	B	Medium	6201
9	741	16-04-2022	B	Medium	8177
10	785	16-04-2022	BB	Medium	1158114
11	1023	16-04-2022	BBB	High	824142
12	891	16-04-2022	BB	Medium	320193
13	936	11-06-2022	BBB	High	943034
15	605	20-06-2022	B	Medium	1844817

	Start Price	End Price	Rate of Change (%)
0	187.491135	136.086594	-27.417052
1	56.471729	42.734833	-24.325263
2	390.305511	502.428802	28.727058
3	198.279999	124.199997	-37.361308
4	88.07	117.639999	33.575565
5	60.558395	38.058285	-37.154404
6	173.071594	128.059479	-26.007801
7	4.19	1.48	-64.677804
8	23.860001	18.25	-23.512156
9	3.879058	3.051214	-21.341387
10	8.82	3.65	-58.616778
11	46.315861	36.614571	-20.945935
12	120.588425	172.203995	42.803088
13	3.047645	4.370698	43.412329
15	9.75	9.82	0.717946

[15 rows x 24 columns]

```
In [211... #Extract S&P 500 rate
sp500_data = yf.download('^GSPC', start='2021-04-01',end='2022-04-01')
sp500_df = pd.DataFrame({
    'Date': sp500_data.index,
    'Start Price': sp500_data['Open'],
    'End Price': sp500_data['Close'],
    'Rate of Change (%)': ((sp500_data['Close'] - sp500_data['Open']) / sp500_data['Open']) * 100
```

```
}
sp500_df.set_index('Date', inplace=True)
print(sp500_df)
```

[*****100%*****] 1 of 1 completed

	Start Price	End Price	Rate of Change (%)
Date			
2021-04-01	3992.780029	4019.870117	0.678477
2021-04-05	4034.439941	4077.909912	1.077472
2021-04-06	4075.570068	4073.939941	-0.039998
2021-04-07	4074.290039	4079.949951	0.138918
2021-04-08	4089.949951	4097.169922	0.176530
...
2022-03-25	4522.910156	4543.060059	0.445507
2022-03-28	4541.089844	4575.520020	0.758192
2022-03-29	4602.859863	4631.600098	0.624400
2022-03-30	4624.200195	4602.450195	-0.470352
2022-03-31	4599.020020	4530.410156	-1.491837

[253 rows x 3 columns]

Data Description

We have two main datasets for our analysis-ready data.

We are using data from a Kaggle csv, and joining this to the yfinance library. First, we downloaded the ESG Kaggle csv, where each row corresponds to a different publicly traded company; this also contains ESG metrics for each company. We have filtered this dataset to only include companies that are traded in USD, and the columns all refer to a specific aspect of the company, such as industry, name, stock close/open/high/low prices, and ESG rating indexes. Then, we join this data to the yfinance library (Yahoo Finance) - from yfinance, we take the stock closing price of the company on 4/1/21 and 4/1/22. We also created a % Change variable, which measures the percentage change in the company stock closing price over the 2 years. All of this information is included in our large dataset - so the columns are company stock ticker, company name, currency, exchange, industry, logo, web url, environment grade, environment level, social grade, governance grade, governance level, environmental score, social score, governance score, total score, last processing date, total grade, total

level, central index key, stock closing price in 4/1/21, stock closing price in 4/1/22, and the percent change between these. Our smaller dataset is meant to be a 'sample' of the original dataset with 722 rows, and involves a few companies that we hand-picked by notability. It contains all of the same columns, but only contains 23 companies.

Our dataframes were created to show different stock prices of various companies over different date ranges, and the original ESG company data was created to compile ESG information for 700 mid / large-cap companies across various industries. Our combined dataframes was made to contrast both stock prices and ESG ratings of companies, and explore any associations. The original ESG company dataset was 'funded' by the efforts of Kaggle user Alistair King, and the yfinance dataset was created by Ran Aroussi as a way around the recent-ish Yahoo Finance API deprecation. For the ESG rating dataset, only mid/large-cap companies are included, so this influences the specific companies that are present in the dataset (the data that was observed and recorded) -- smaller companies will not be 'observed' here.

The preprocessing was described above; we filtered our 722 row dataset for NaNs for our large dataset, and then for our sample dataset, we filtered all 722 companies down to 23 of interest, and joined all data to the yfinance library. Specifically, for each company, we acquired the stock closing price for 4/1/21 and 4/1/24, then created a stock change percentage variable between these two dates.

Individuals are not involved in the data directly, as each observation corresponds to an entire company.

Our raw source data can be found in the yfinance library and <https://www.kaggle.com/datasets/alistaiking/public-company-esg-ratings-dataset/data>, and the specific csv is here: https://github.com/phoebewang28/info-2950-project/blob/main/esg_data.csv.

Data Limitations

1. For our smaller dataset, the current sample of 23 companies was chosen manually by us as we wanted to get a range of industries that are well-known. This is limited to the companies that only use USD, and again, this data only records large/mid-cap companies, so this selection may not be fully representative of all US companies with ESG ratings. Depending on the results of our analyses, we may consider random sampling or expanding the sample size to improve representativeness of our sample dataset.
2. We are currently comparing the rate of change of the sample stocks to the S&P 500, but other measures of stock

performance might provide more valuable insights. For now, we are focusing on the rate of change between the closing prices of 4/1/21 and 4/1/22.

3. Some stock data from yfinance is missing date information, which causes missing values when extracting prices. One company in our sample had this issue, so we had to exclude it to ensure consistency.
4. Since we are exploring potential connections between ESG ratings and company stock performance, we may need to sample not only by industry but also by ESG rating levels to ensure a more balanced and comprehensive analysis of the different ESG performance tiers (for our sample dataset)
5. ESG is a constant value retrieved from different days for each company in the month of April, 2022, while stock prices for these companies changes over time. We're unable to perform time-series analysis on ESG rating and stock informations due to the fact.
6. ESG is evaluated annually, which might not be accurately tracking the actual environmental performance of the company. Thus, when considering short-term impact of the company's esg policies, it's likely for that policy change to affect stock but not reflected on company's ESG rating.

Exploratory Data Analysis

part one - exploring different average environmental, social, governance, and total ESG scores by industry

```
In [212... average_total_score_by_industry = esg.groupby('industry')['total_score'].mean().reset_index()
average_total_score_by_industry.columns = ['Industry', 'Average Total ESG Score']
print(average_total_score_by_industry.head(5))

avg_totals = average_total_score_by_industry.sort_values(by = 'Average Total ESG Score', ascending = False)
print("best average ESG scores")
print(avg_totals.head(5))

average_total_score_by_industry = average_total_score_by_industry.sort_values(by = 'Average Total ESG Score', ascending = True)
print("worst average ESG scores")
```

```
print(average_total_score_by_industry.head(5))
```

	Industry	Average Total ESG Score
0	Aerospace & Defense	633.000000
1	Aerospace and Defense	1089.700000
2	Airlines	838.714286
3	Auto Components	950.000000
4	Automobiles	859.250000

best average ESG scores

	Industry	Average Total ESG Score
46	Utilities	1190.633333
44	Tobacco	1165.000000
24	Industrial Conglomerates	1157.000000
34	Packaging	1146.666667
20	Food Products	1131.071429

worst average ESG scores

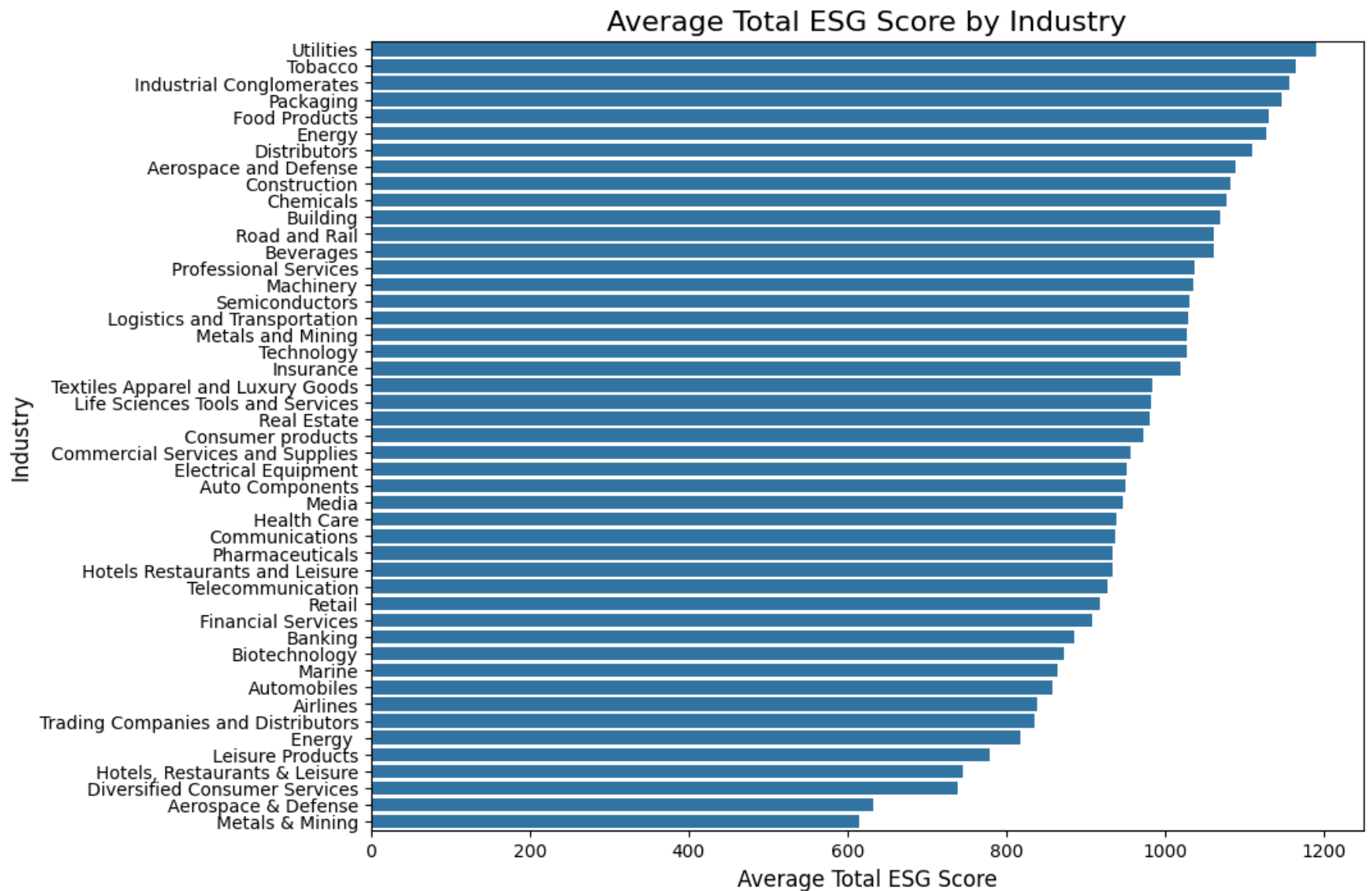
	Industry	Average Total ESG Score
32	Metals & Mining	615.000000
0	Aerospace & Defense	633.000000
15	Diversified Consumer Services	739.333333
23	Hotels, Restaurants & Leisure	746.000000
26	Leisure Products	780.000000

```
In [213... # Bar chart of Toatl ESG score by Industry
average_total_score_by_industry = average_total_score_by_industry.sort_values('Average Total ESG Score',

# Create the horizontal bar chart
plt.figure(figsize=(10, 8))
sns.barplot(x='Average Total ESG Score', y='Industry', data=average_total_score_by_industry)

# Add labels and title
plt.title('Average Total ESG Score by Industry', fontsize=16)
plt.xlabel('Average Total ESG Score', fontsize=12)
plt.ylabel('Industry', fontsize=12)
```

```
Out[213... Text(0, 0.5, 'Industry')
```



interestingly (and somewhat predictably) - the industries with the lowest ESG scores are Metals & Mining, Aerospace & Defense, Diversified Consumer Services, Hotels, Restaurants & Leisure, Leisure Products, Auto Components, Airlines, and Automobiles. the industries with the highest ESG scores are Utilities, Tobacco, Industrial Conglomerates, Packaging, and Energy.

future steps: sort different industries by just Environmental score, just Social score, and just Governance score to see if

these differ significantly.

```
In [214... print(esg.columns)
```

```
Index(['ticker', 'name', 'currency', 'exchange', 'industry', 'logo', 'weburl',  
      'environment_grade', 'environment_level', 'social_grade',  
      'social_level', 'governance_grade', 'governance_level',  
      'environment_score', 'social_score', 'governance_score', 'total_score',  
      'last_processing_date', 'total_grade', 'total_level', 'cik',  
      'Start Price', 'End Price', 'Rate of Change (%)'],  
      dtype='object')
```

```
In [215... avg_environmental = esg.groupby('industry')['environment_score'].mean().reset_index()  
avg_environmental.columns = ['Industry', 'Average Environmental Score']  
print(avg_environmental.head(5))  
  
avg_enviro = avg_environmental.sort_values(by = 'Average Environmental Score', ascending = False)  
print("best environmental scores")  
print(avg_enviro.head(5))  
  
avg_enviro = avg_environmental.sort_values(by = 'Average Environmental Score', ascending = True)  
print("worse environmental scores")  
print(avg_enviro.head(5))
```


	Industry	Average Environmental Score
0	Aerospace & Defense	225.000000
1	Aerospace and Defense	485.000000
2	Airlines	326.714286
3	Auto Components	407.500000
4	Automobiles	393.000000

best environmental scores

	Industry	Average Environmental Score
46	Utilities	550.966667
24	Industrial Conglomerates	534.666667
34	Packaging	530.000000
14	Distributors	525.000000
17	Energy	516.611111

worse environmental scores

	Industry	Average Environmental Score
32	Metals & Mining	215.000000
0	Aerospace & Defense	225.000000
26	Leisure Products	252.333333
23	Hotels, Restaurants & Leisure	270.000000
15	Diversified Consumer Services	286.000000

```
In [216... avg_social = esg.groupby('industry')['social_score'].mean().reset_index()
avg_social.columns = ['Industry', 'Average Social Score']
print(avg_social.head(5))

avg_social = avg_social.sort_values(by = 'Average Social Score', ascending = False)
print("best social scores")
print(avg_social.head(5))

avg_social = avg_social.sort_values(by = 'Average Social Score', ascending = True)
print("worse social scores")
print(avg_social.head(5))
```

	Industry	Average Social Score
0	Aerospace & Defense	203.000000
1	Aerospace and Defense	305.700000
2	Airlines	265.714286
3	Auto Components	272.500000
4	Automobiles	229.250000

best social scores

	Industry	Average Social Score
46	Utilities	357.400000
39	Road and Rail	329.400000
44	Tobacco	327.000000
34	Packaging	322.666667
24	Industrial Conglomerates	318.000000

worse social scores

	Industry	Average Social Score
32	Metals & Mining	200.000000
0	Aerospace & Defense	203.000000
4	Automobiles	229.250000
15	Diversified Consumer Services	233.333333
23	Hotels, Restaurants & Leisure	248.500000

```
In [217... avg_governance = esg.groupby('industry')['governance_score'].mean().reset_index()
avg_governance.columns = ['Industry', 'Average Governance Score']
print(avg_governance.head(5))

avg_governance = avg_governance.sort_values(by = 'Average Governance Score', ascending = False)
print("best governance scores")
print(avg_governance.head(5))

avg_governance = avg_governance.sort_values(by = 'Average Governance Score', ascending = True)
print("worse governance scores")
print(avg_governance.head(5))
```

	Industry	Average Governance Score
0	Aerospace & Defense	205.000000
1	Aerospace and Defense	299.000000
2	Airlines	246.285714
3	Auto Components	270.000000
4	Automobiles	237.000000

best governance scores

	Industry	Average Governance Score
44	Tobacco	328.000000
14	Distributors	307.666667
24	Industrial Conglomerates	304.333333
20	Food Products	303.214286
36	Professional Services	300.500000

worse governance scores

	Industry	Average Governance Score
32	Metals & Mining	200.0
0	Aerospace & Defense	205.0
15	Diversified Consumer Services	220.0
23	Hotels, Restaurants & Leisure	227.5
4	Automobiles	237.0

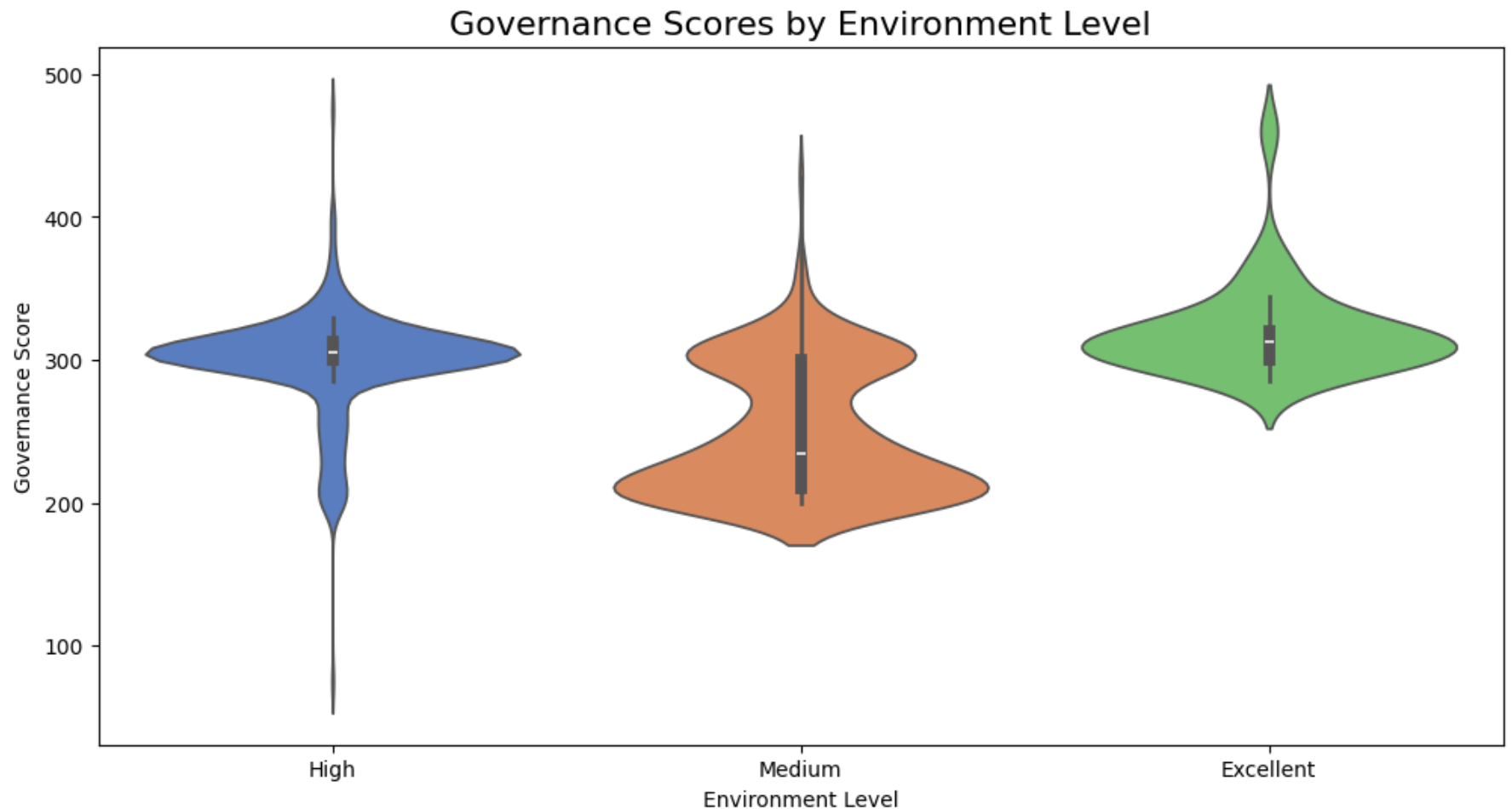
part two

```
In [218... #Violin Plot: Governance Score by Environment Level
plt.figure(figsize=(12, 6))
sns.violinplot(data=esg, x='environment_level', y='governance_score', palette='muted')
plt.title('Governance Scores by Environment Level', fontsize=16)
plt.xlabel('Environment Level')
plt.ylabel('Governance Score')
plt.show()
```

/var/folders/j2/kdbytjmd57914f_8rxw949240000gn/T/ipykernel_30230/2624109517.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.violinplot(data=esg, x='environment_level', y='governance_score', palette='muted')
```



```
In [219... #violin graph
print(relevant_esg.columns)
correlation = relevant_esg['total_score'].corr(relevant_esg['Rate of Change (%)'])
print(f'Correlation between Total ESG Score and Rate of Change (%): {correlation}')

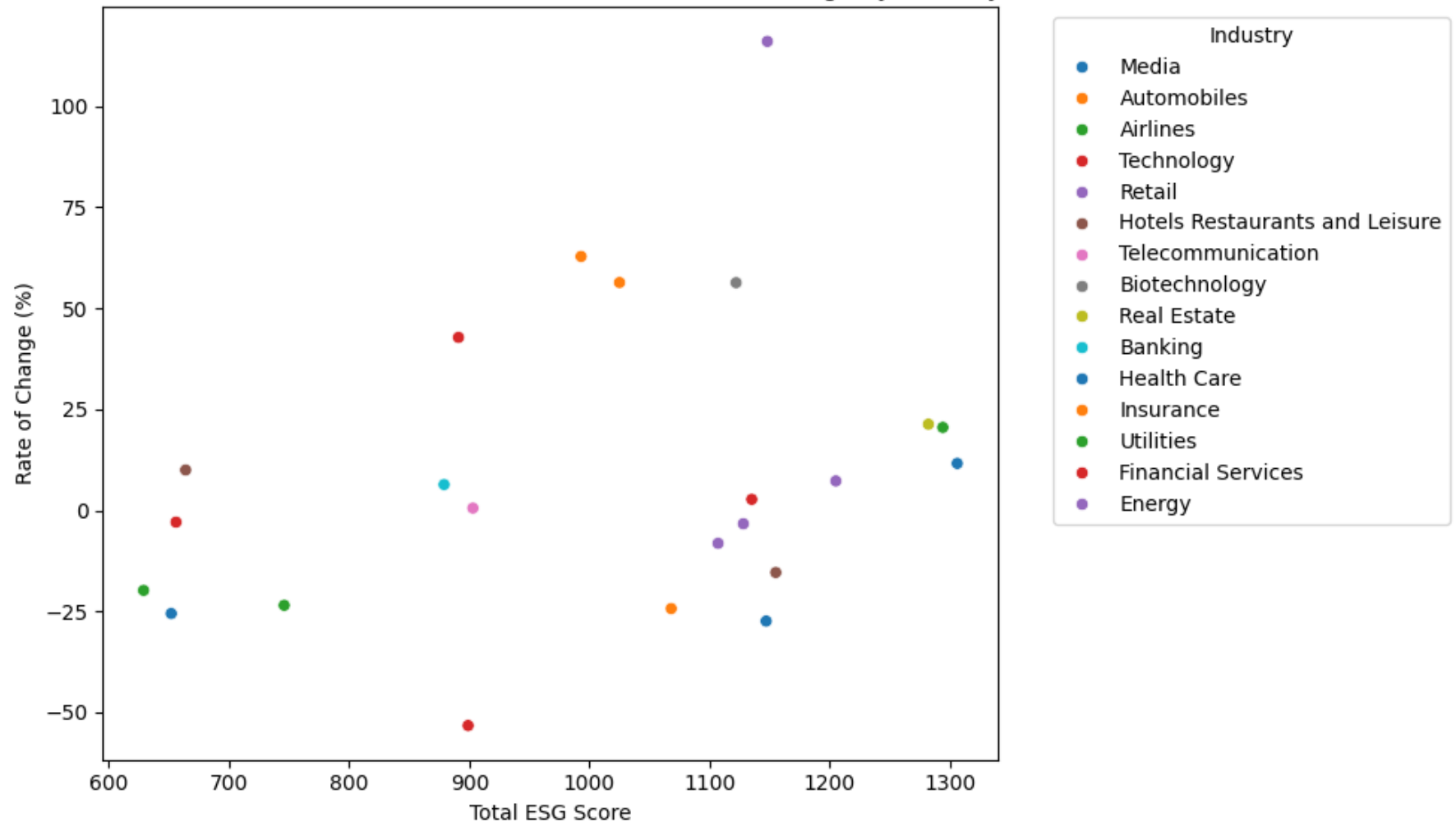
plt.figure(figsize=(10, 6))
sns.scatterplot(data=relevant_esg, x='total_score', y='Rate of Change (%)', hue='industry', palette='tab10')
plt.title('Correlation Between Total ESG Score and Stock Price Change by Industry')
plt.xlabel('Total ESG Score')
plt.ylabel('Rate of Change (%)')
plt.legend(title='Industry', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.tight_layout()
```

```
plt.show()
```

```
Index(['ticker', 'name', 'currency', 'exchange', 'industry', 'logo', 'weburl',  
      'environment_grade', 'environment_level', 'social_grade',  
      'social_level', 'governance_grade', 'governance_level',  
      'environment_score', 'social_score', 'governance_score', 'total_score',  
      'last_processing_date', 'total_grade', 'total_level', 'cik',  
      'Start Price', 'End Price', 'Rate of Change (%)'],  
      dtype='object')
```

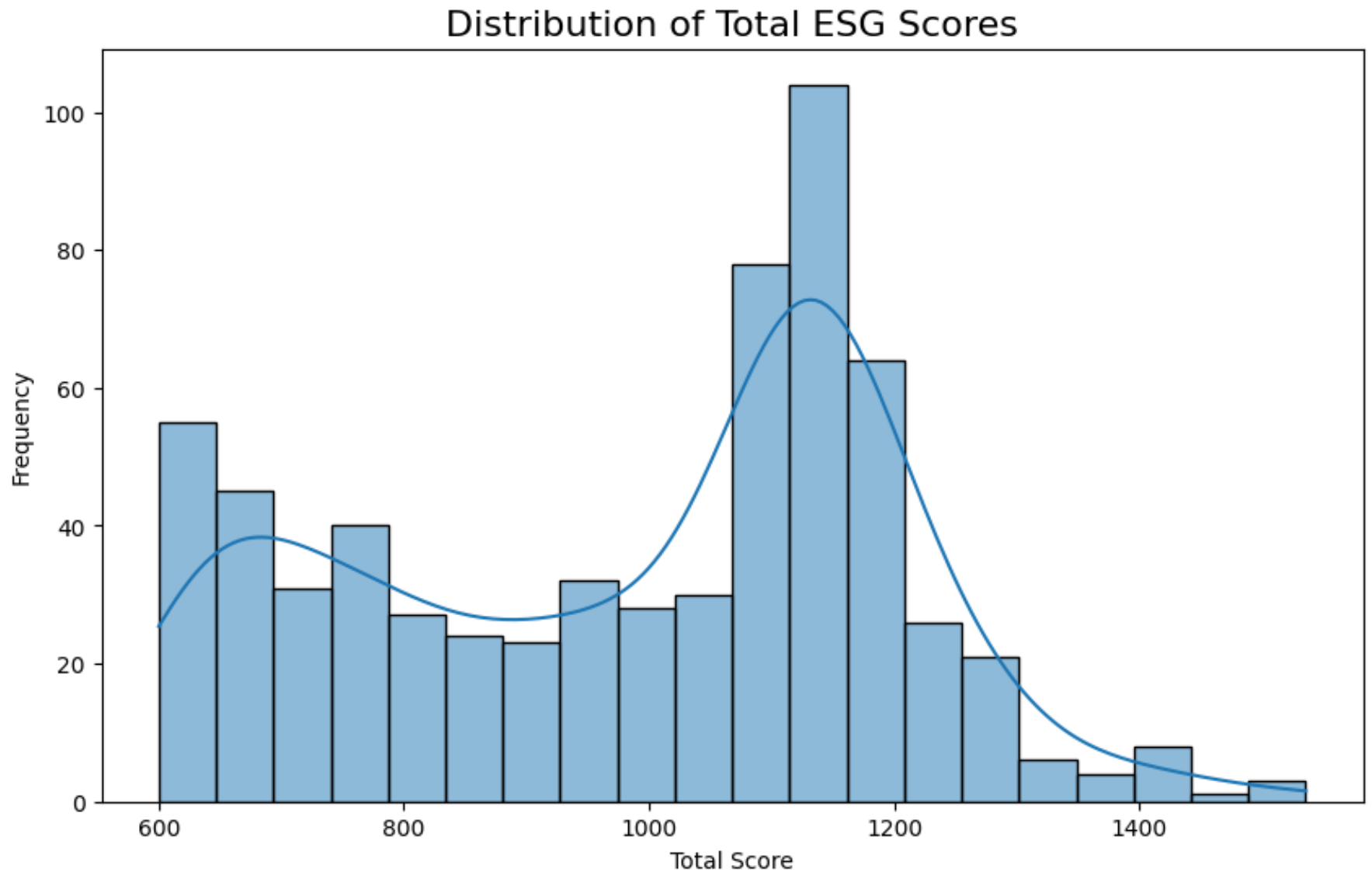
Correlation between Total ESG Score and Rate of Change (%): 0.2933148427717424

Correlation Between Total ESG Score and Stock Price Change by Industry



In [220...

```
# Plot the distribution of total_score
plt.figure(figsize=(10, 6))
sns.histplot(esg['total_score'], kde=True, bins=20)
plt.title('Distribution of Total ESG Scores', fontsize=16)
plt.xlabel('Total Score')
plt.ylabel('Frequency')
plt.show()
```

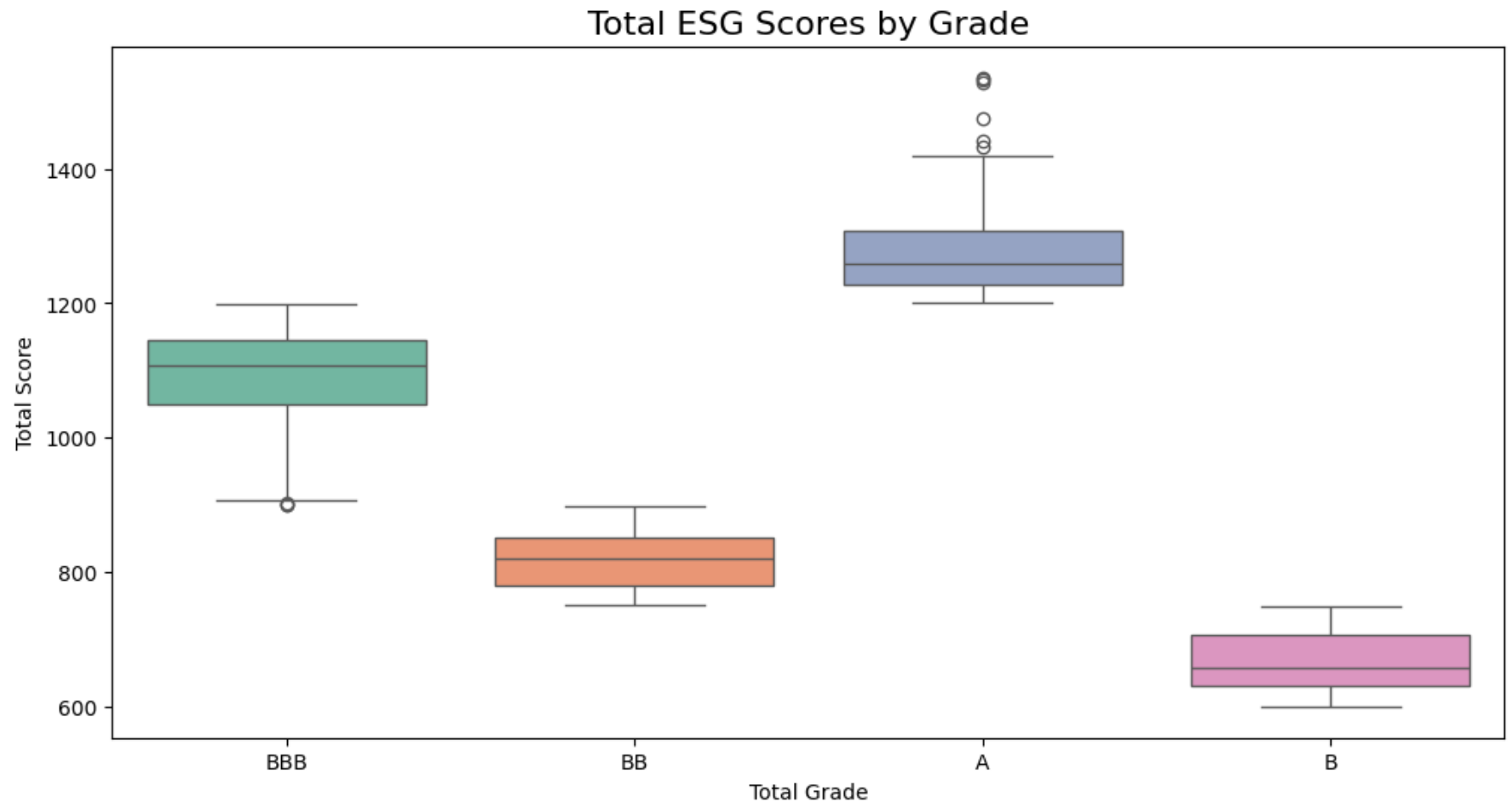


```
In [221... #Boxplot of Total Scores by Total Grade
plt.figure(figsize=(12, 6))
sns.boxplot(data=esg, x='total_grade', y='total_score', palette='Set2')
plt.title('Total ESG Scores by Grade', fontsize=16)
plt.xlabel('Total Grade')
plt.ylabel('Total Score')
plt.show()
```

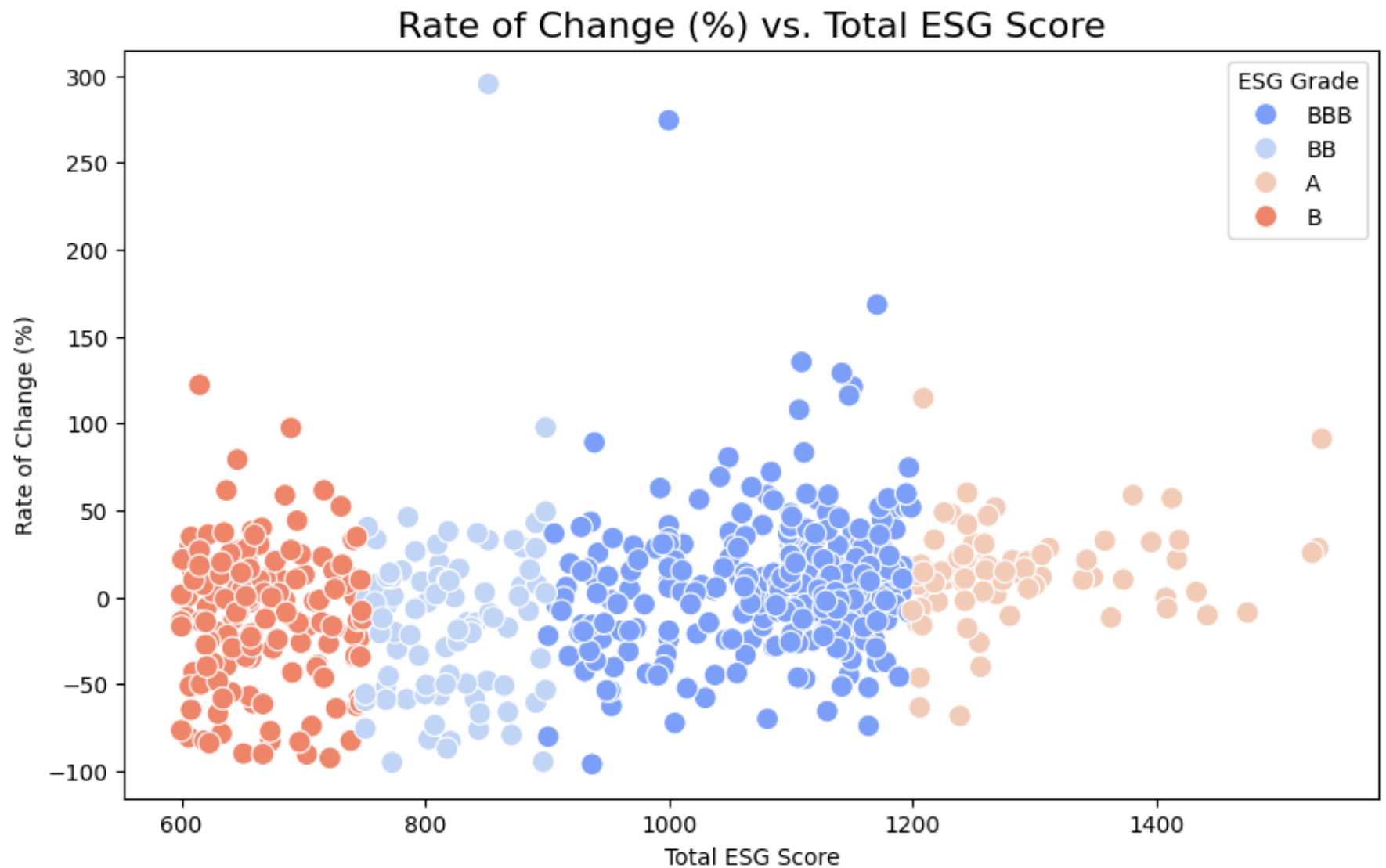
/var/folders/j2/kdbytjmd57914f_8rxw949240000gn/T/ipykernel_30230/3588776245.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.boxplot(data=esg, x='total_grade', y='total_score', palette='Set2')
```



```
In [222... #Scatter Plot: Rate of Change vs. Total ESG Score
plt.figure(figsize=(10, 6))
sns.scatterplot(data=esg, x='total_score', y='Rate of Change (%)', hue='total_grade', palette='coolwarm',
plt.title('Rate of Change (%) vs. Total ESG Score', fontsize=16)
plt.xlabel('Total ESG Score')
plt.ylabel('Rate of Change (%)')
plt.legend(title="ESG Grade", loc='upper right')
plt.show()
```

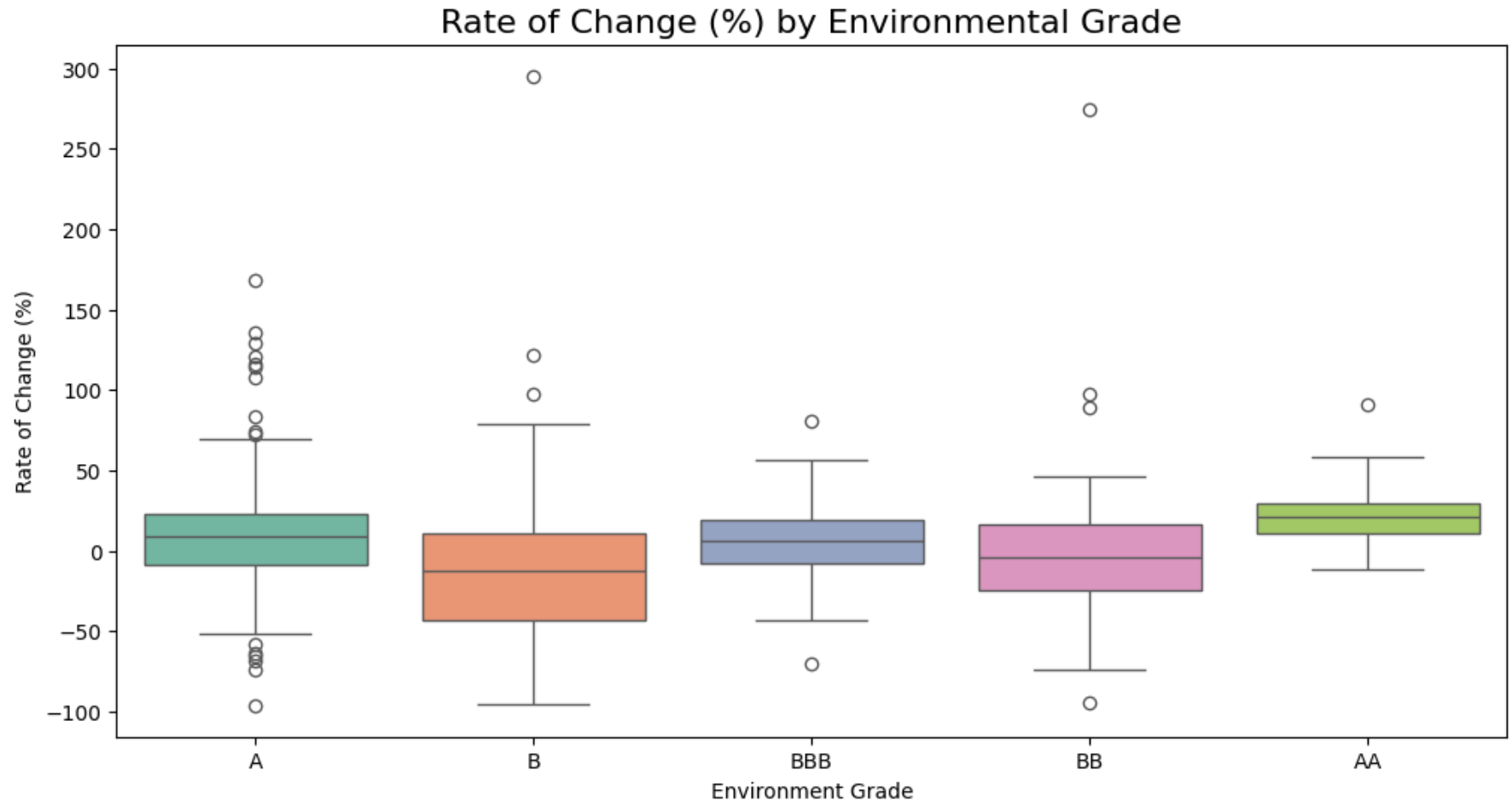



```
In [223... #Boxplot: Rate of Change by Environment Grade
plt.figure(figsize=(12, 6))
sns.boxplot(data=esg, x='environment_grade', y='Rate of Change (%)', palette='Set2')
plt.title('Rate of Change (%) by Environmental Grade', fontsize=16)
plt.xlabel('Environment Grade')
plt.ylabel('Rate of Change (%)')
plt.show()
```

```
/var/folders/j2/kdbytjmd57914f_8rxw949240000gn/T/ipykernel_30230/3614906153.py:3: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.boxplot(data=esg, x='environment_grade', y='Rate of Change (%)', palette='Set2')
```



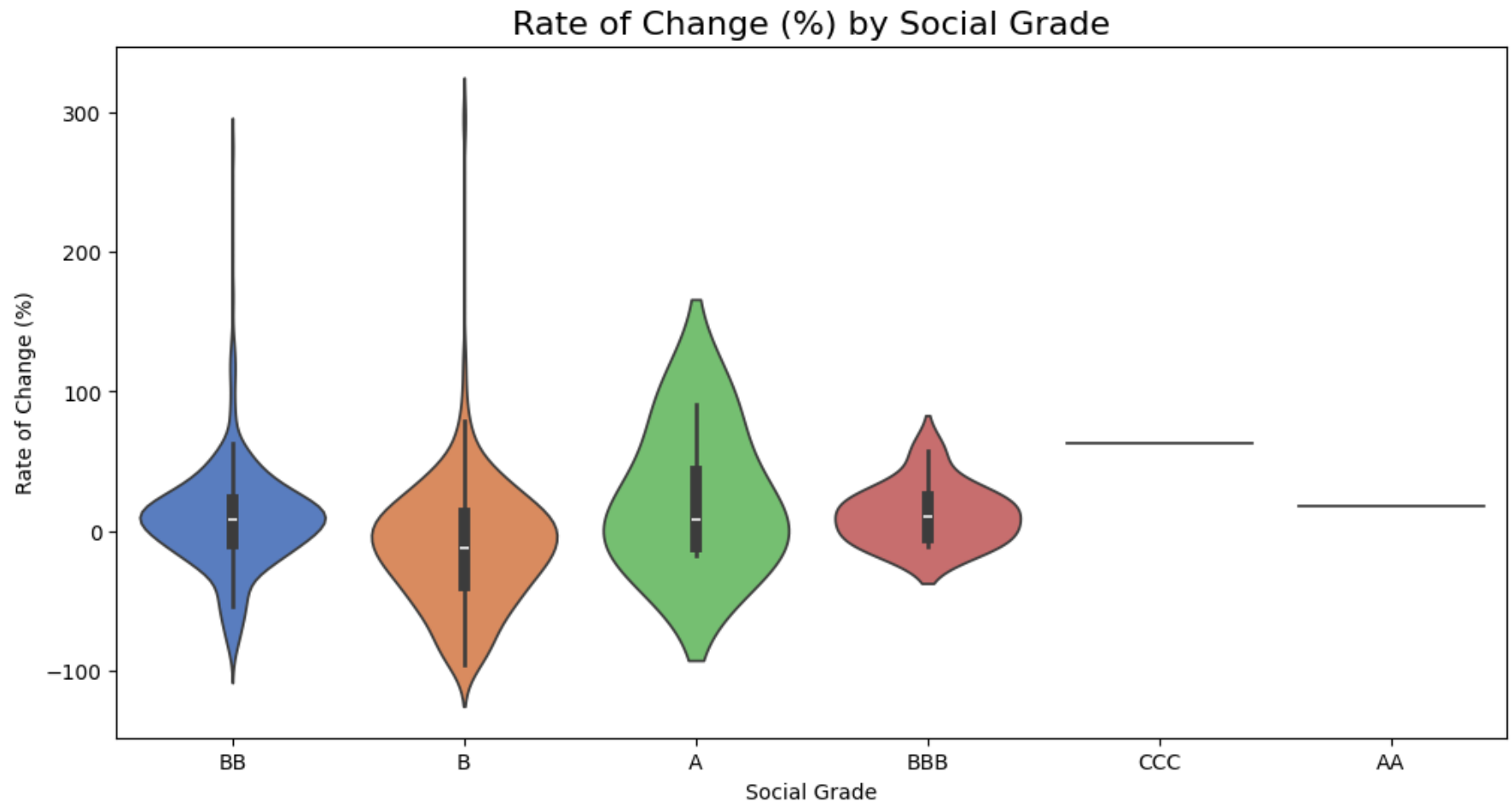
```
In [224... #Violin Plot: Rate of Change by Social Grade
plt.figure(figsize=(12, 6))
sns.violinplot(data=esg, x='social_grade', y='Rate of Change (%)', palette='muted')
plt.title('Rate of Change (%) by Social Grade', fontsize=16)
plt.xlabel('Social Grade')
plt.ylabel('Rate of Change (%)')
```

```
plt.show()
```

/var/folders/j2/kdbytjmd57914f_8rxw949240000gn/T/ipykernel_30230/4258631131.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.violinplot(data=esg, x='social_grade', y='Rate of Change (%)', palette='muted')
```



```
In [225... correlation_matrix=esg[['environment_score', 'social_score', 'governance_score', 'Rate of Change (%)']].c
plt.figure(figsize=(8, 6))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', linewidths=0.5)
plt.title('Correlation Matrix Between ESG Subscores and Stock Price Change')
plt.xticks(rotation=45)
```

```
plt.show()
```

```
correlation_environment = esg['environment_score'].corr(esg['Rate of Change (%)'])
```

```
correlation_social = esg['social_score'].corr(esg['Rate of Change (%)'])
```

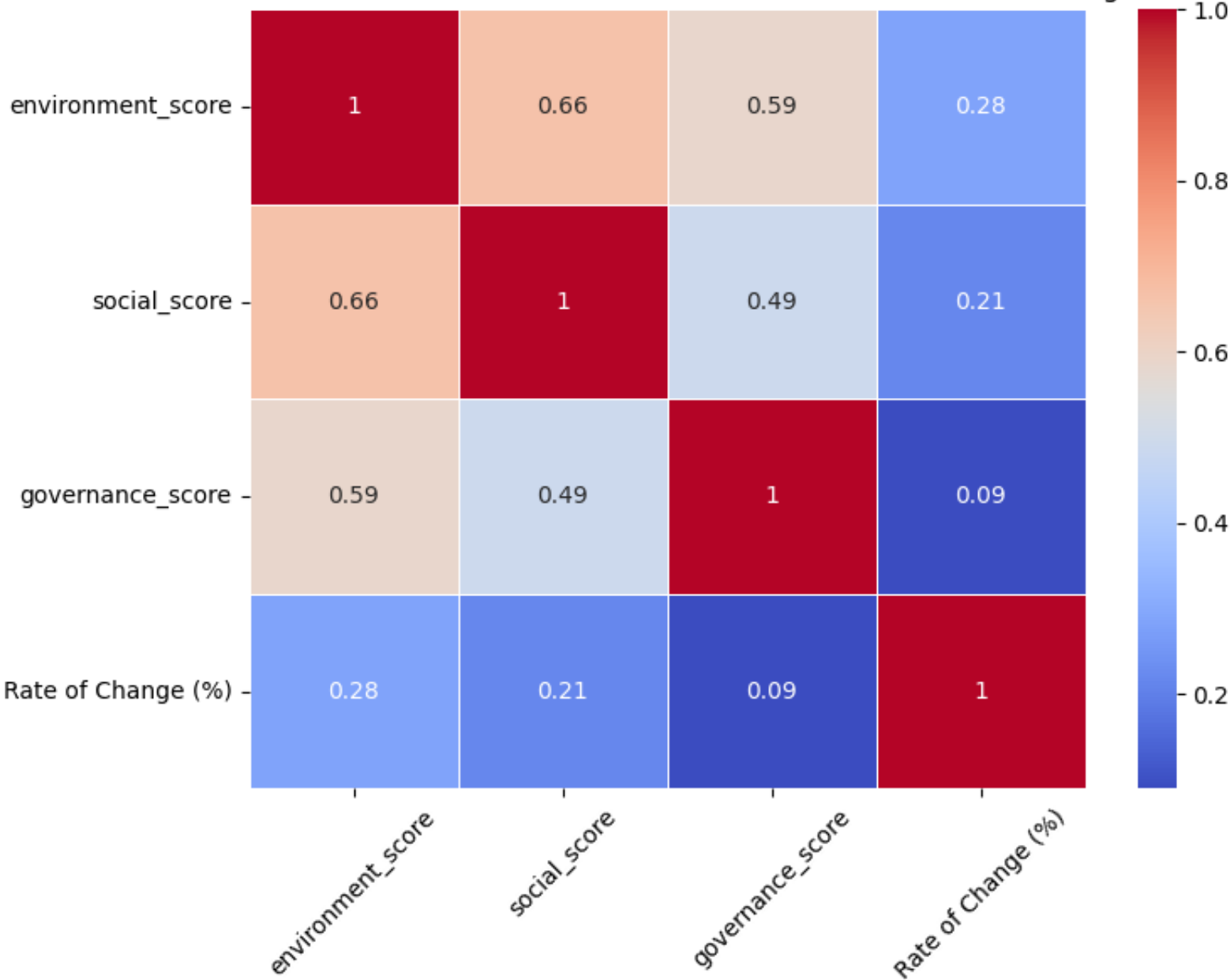
```
correlation_governance = esg['governance_score'].corr(esg['Rate of Change (%)'])
```

```
print(f'Correlation between Environmental Score and Rate of Change (%): {correlation_environment:.3f}')
```

```
print(f'Correlation between Social Score and Rate of Change (%): {correlation_social:.3f}')
```

```
print(f'Correlation between Governance Score and Rate of Change (%): {correlation_governance:.3f}')
```

Correlation Matrix Between ESG Subscores and Stock Price Change



Correlation between Environmental Score and Rate of Change (%): 0.283
Correlation between Social Score and Rate of Change (%): 0.215
Correlation between Governance Score and Rate of Change (%): 0.090

In [226...

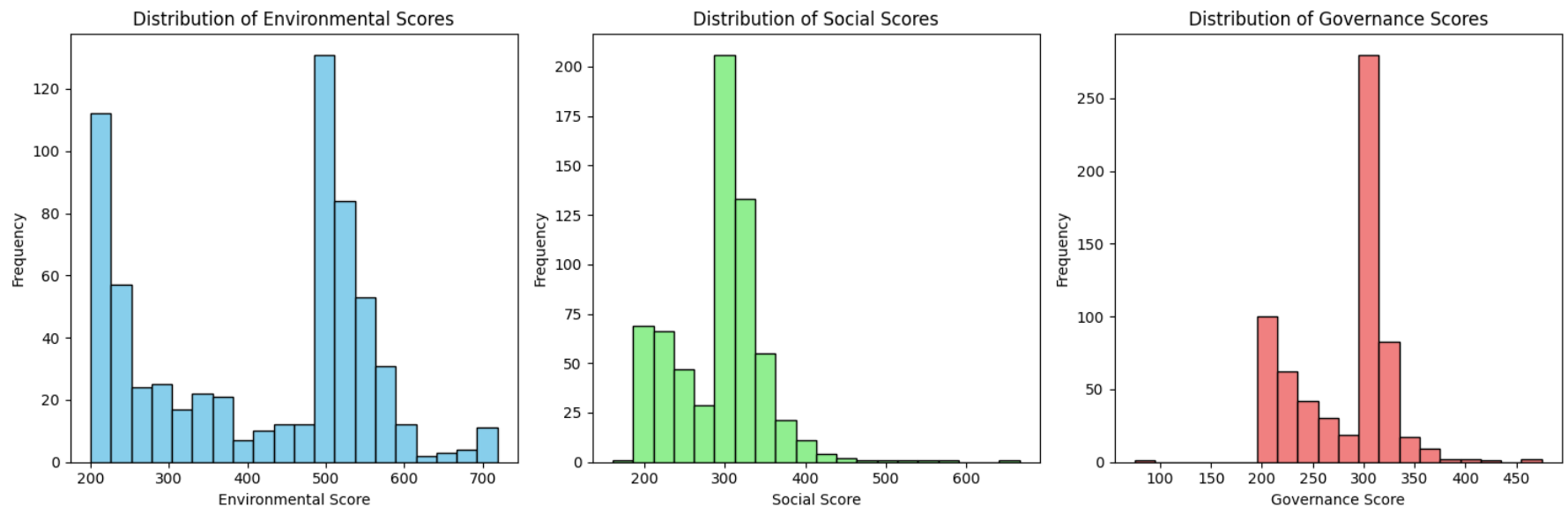
```
# Plot histograms for each ESG subscore
plt.figure(figsize=(15, 5))

plt.subplot(1, 3, 1)
plt.hist(esg['environment_score'].dropna(), bins=20, color='skyblue', edgecolor='black')
plt.xlabel('Environmental Score')
plt.ylabel('Frequency')
plt.title('Distribution of Environmental Scores')

plt.subplot(1, 3, 2)
plt.hist(esg['social_score'].dropna(), bins=20, color='lightgreen', edgecolor='black')
plt.xlabel('Social Score')
plt.ylabel('Frequency')
plt.title('Distribution of Social Scores')

plt.subplot(1, 3, 3)
plt.hist(esg['governance_score'].dropna(), bins=20, color='lightcoral', edgecolor='black')
plt.xlabel('Governance Score')
plt.ylabel('Frequency')
plt.title('Distribution of Governance Scores')

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

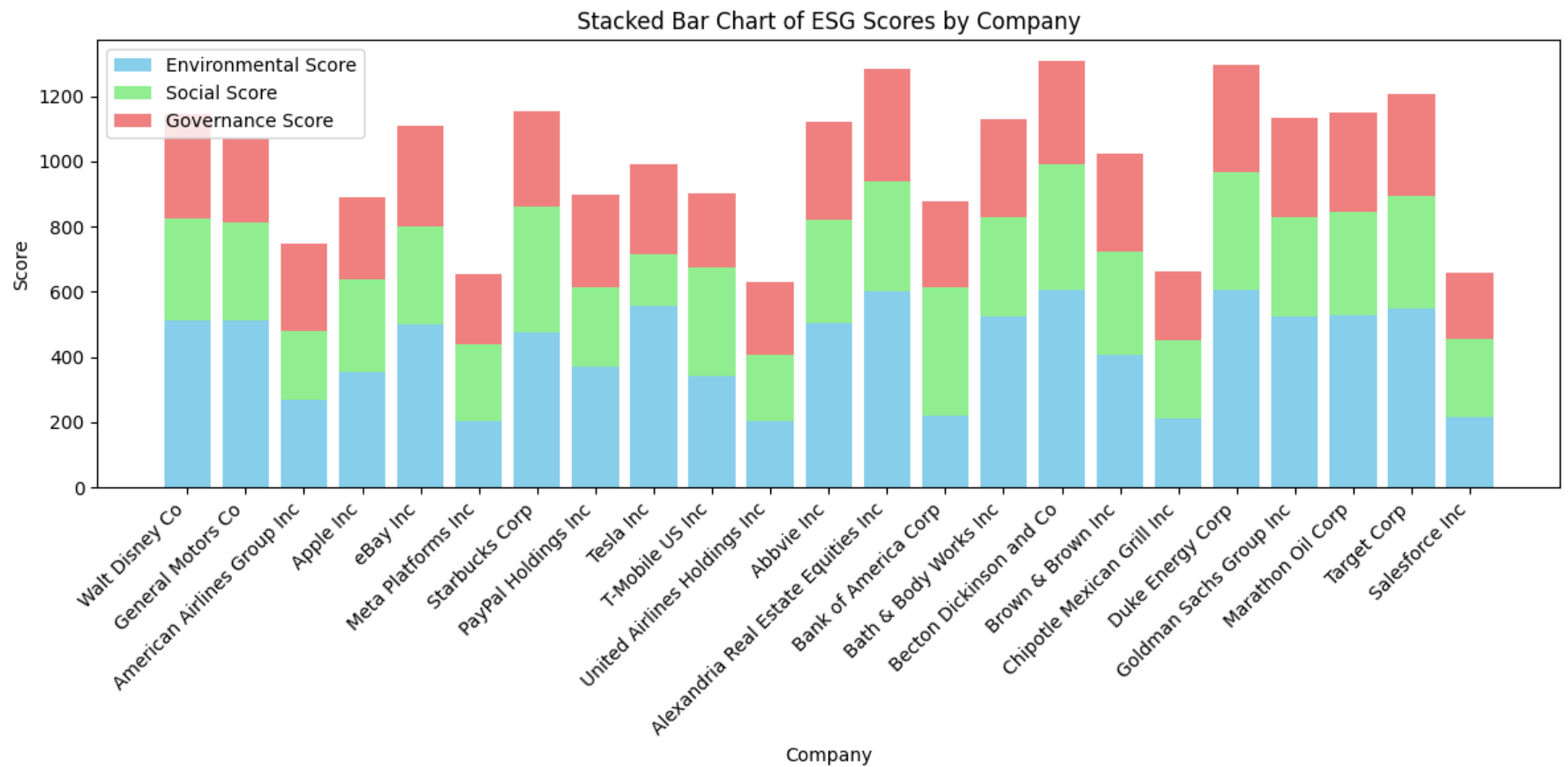


```
In [227... env_scores = relevant_esg['environment_score'].values
soc_scores = relevant_esg['social_score'].values
gov_scores = relevant_esg['governance_score'].values
filtered_companies = relevant_esg['name'].values

x = np.arange(len(filtered_companies))

plt.figure(figsize=(12, 6))
plt.bar(x, env_scores, color='skyblue', label='Environmental Score')
plt.bar(x, soc_scores, bottom=env_scores, color='lightgreen', label='Social Score')
plt.bar(x, gov_scores, bottom=env_scores + soc_scores, color='lightcoral', label='Governance Score')

plt.xticks(x, filtered_companies, rotation=45, ha='right')
plt.xlabel('Company')
plt.ylabel('Score')
plt.title('Stacked Bar Chart of ESG Scores by Company')
plt.legend()
plt.tight_layout()
plt.show()
```

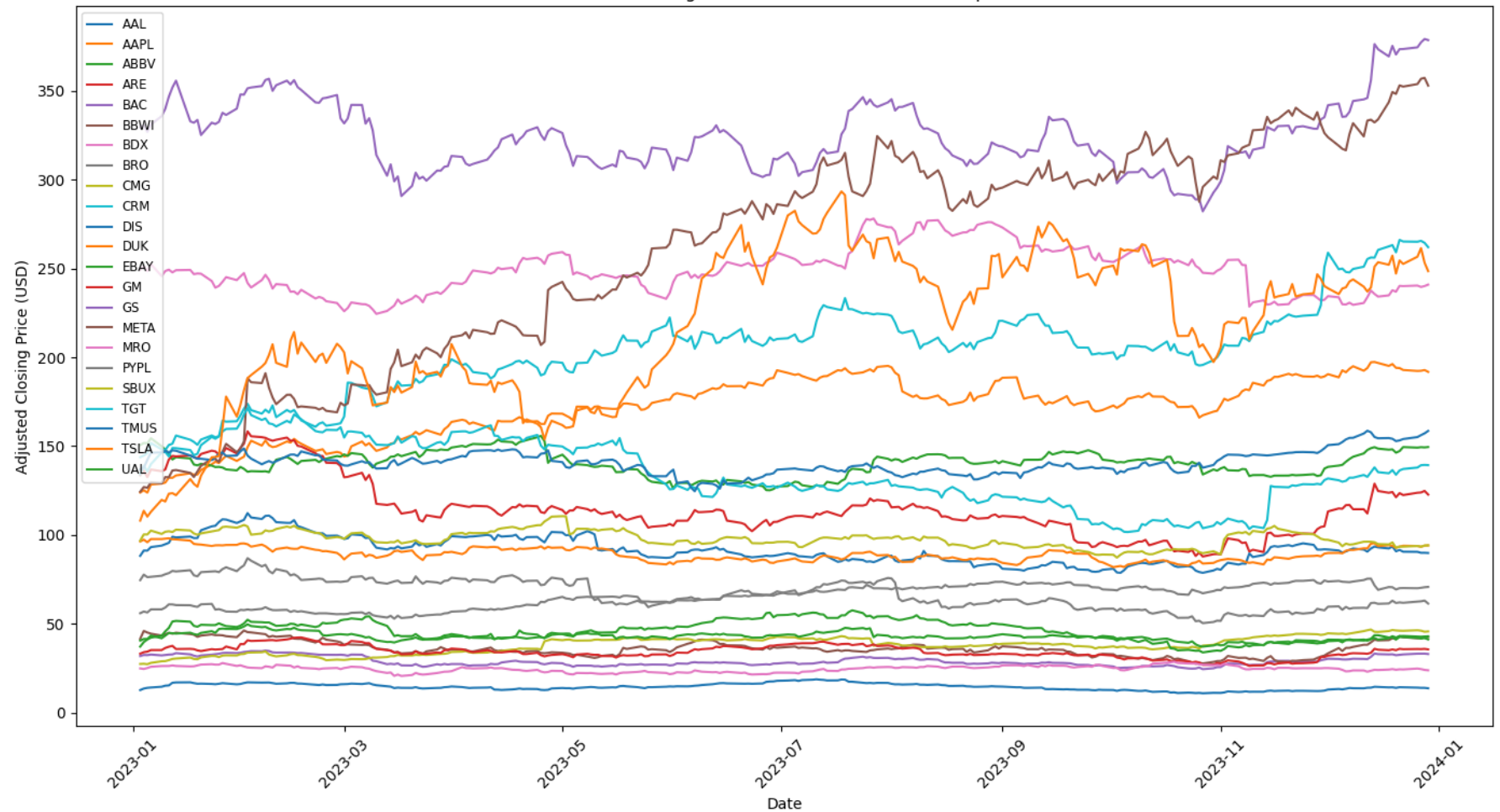


```
In [228... data = yf.download(list(relevant_esg['ticker']), start='2023-01-01', end='2024-01-01')['Adj Close']
```

```
plt.figure(figsize=(14, 8))
for ticker in data.columns:
    plt.plot(data.index, data[ticker], label=ticker)

plt.xlabel('Date')
plt.ylabel('Adjusted Closing Price (USD)')
plt.title('Stock Price Changes Over Time for Selected Companies')
plt.legend(loc='upper left', fontsize='small')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```


Stock Price Changes Over Time for Selected Companies



```
In [229... X = relevant_esg[['environment_score', 'social_score', 'governance_score']] # Independent variables
y = relevant_esg['Rate of Change (%)'] # Dependent variable

model = LinearRegression().fit(X, y)
print(f"Environmental Score Coefficient: {model.coef_[0]}")
print(f"Social Score Coefficient: {model.coef_[1]}")
print(f"Governance Score Coefficient: {model.coef_[2]}")
print(f"Intercept: {model.intercept_}")
```

Environmental Score Coefficient: 0.1079341804300019
Social Score Coefficient: -0.023698725388349346
Governance Score Coefficient: -0.07501676055180559
Intercept: -8.825036870813836

Regression Coefficients Interpretation

For every 1-unit increase in the Environmental score (assuming all other factors remain constant), the stock return is expected to increase by 0.1079 units. The positive coefficient suggests that higher Environmental scores are associated with better stock performance (or higher returns).

For every 1-unit increase in the Social score (with other variables constant), the stock return is expected to decrease by 0.0237 units. The negative coefficient indicates that better Social scores might be associated with lower stock performance, but generally, since the coefficient is so close to 0, Social scores seem to have little impact on stock performance.

For every 1-unit increase in the Governance score (with other variables constant), stock return is expected to decrease by 0.0750 units. This negative coefficient suggests that improvements in governance (e.g., stricter regulation or more ethical practices) are associated with slightly lower stock returns. This could imply that governance improvements come at a financial cost.

```
In [230... esg_corr = relevant_esg.loc[:, ['environment_score', 'social_score',  
                                'governance_score']]  
print(esg_corr.corr(numeric_only=True))
```

	environment_score	social_score	governance_score
environment_score	1.000000	0.443407	0.866404
social_score	0.443407	1.000000	0.548976
governance_score	0.866404	0.548976	1.000000

```
In [231... print(esg_corr.cov(numeric_only=True))
```

	environment_score	social_score	governance_score
environment_score	20758.632411	3947.505929	5092.833992
social_score	3947.505929	3818.059289	1383.930830
governance_score	5092.833992	1383.930830	1664.482213

Correlation and Covariance

Let's observe the correlation/covariance between the different ESG scores themselves! All the correlations are positive, meaning that there is positive correlation between the scores (when one increases, the other does too). Governance and Environment have a particularly strong correlation, suggesting that companies who invest in environmental factors likely also care about governance (or perhaps some government regulations align with environmental issues). On the other hand, social factors seem to have just a moderately positive relationship with both the other variables.

While the covariance values agree with these claims, it's interesting to note how large the variance is for Environmental Scores (20758.63). This suggests that there is a large spread in environmental performance among the companies we chose, while governance scores are much more consistent.

Questions for Reviewers

1. After combining our yfinance and ESG data, do we have large enough datasets to satisfy the complexity requirement for the research question?
2. Any advice recommended to follow when we trying to take sample from the population? maybe by industry? how many industry to take sample from? maybe stratify to include all ESG grade? (as a reminder - we have one big dataset with 700+ companies and joined data that we did some EDA with, but we also want to include a smaller sample dataset so we can look at some individual companies as well).
3. How many visualization and statistics are recommended for the final project? (ballpark range would be helpful) Do the visualizations we currently have seem like they're on the right path for the final phases?
4. Regarding the visualizations and chunks we have made for our EDA so far: should we explore these specific visualizations more in depth? OR should we expand our DA to other variables in the datasets that we maybe haven't used yet?