# State Street Global Advisors UConn Senior Design Project

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**Project Goal:** Develop a full-scale model in R that can read various ESG data and calculate a correlation coefficient to stock price performance over time.

#### Project Overview:

- There are fives R files associated with the model.
  - o MSCI to Price Correlation
  - o GICS Sector Average
  - CQ Score to MSCI Correlation
  - o CQ Score to Price Correlation
  - o TruCost to Price Correlation
- There are four ESG data sets being used with this model.
  - Morgan Stanley Capital International (MSCI)
  - o Global Industry Classification Standard (GICS) Sector
  - Content Quality (CQ)
  - o TruCost

#### Project Development:

- R-Studio has been utilized as the IDE for development throughout this project.
- R-Studio Download

### File Overview:

File Name	Functionality	Input(s)	Output(s)
MSCI_Price_correlation.r	Correlates Price to MSCI Data (ESG).	1. MSCI Data (MSCIData_2014_to_2018_2021 _to_2022.csv)	1. Table of Correlations (10_Years_Correlation_Table. csv)
		2. Price Data (PriceData_2012_2022.csv)	2. MSCI and Price Data Merged (MSCI_Price_Data.csv)
GICS_Sector_Avg.r	Calculates the GICS average per sector.	1. Price Data (Price_Data3_ISIN.csv)	1. GICS Sector Averages (GICS_Sector_MSCI_Correl ation Avg.cs)
	Determines outliers.     Correlates GICS Sector Average to MSCI Data.	2. MSCI Correlations (fixed_10_Years_Correlation_Ta ble.csv)  3. GICS Sectors	2. Outliers by GICS Sector Avg (Outliers_By_GICS_Sector.c sv)
		(GICS_Sectors.csv)	31)
CQ Score to ESG MSCI.r	Correlates CQ score to MSCI Data (ESG).	1. CQ Scores Data (cq.csv) 2. Cleaned CQ Data (Removed_CQ_Data.csv) 3. MSCI Data (ESG_Price_Date_Data.csv)	1. CQ Score to MSCI Correlation (CQ_Score_to_ESG_MSCI_ Correlation.csv)
CQ Score to Price.r	Correlates CQ score to Price.	CQ Scores Data (cq.csv)      Cleaned CQ Data     (Removed_CQ_Data.csv)     This includes price data	1. CQ Score to Price Correlation (CQ_Score_to_Price_Correlations.csv)
truCost_correlation.r	Correlates Price to TruCost.	1. TruCost Data (truCost.csv) 2. Price Data (PriceData_2012_2022.csv)	1. TruCost to Price Correlations (TruCost_to_Price_Correlatio n.csv)

### MSCI to Price Correlation Model

#### Part 1 Input:

stock\_price\_data from PriceData\_2012\_2022.csv

DATE_DIM_ID +	BB_TICKER_CD	ISIN_CD	PBD_PRICE_AMT	year_month <sup>‡</sup>
2014-08-29	ZTS US	US98978V1035	35.44	2014-08
2014-09-04	ZTS US	US98978V1035	35.51	2014-09
2014-09-10	ZTS US	US98978V1035	36.24	2014-09
2014-09-12	ZTS US	US98978V1035	36.18	2014-09

#### Part 2

Find the average monthly price per security

ISIN_CD <sup>‡</sup>	year_month	avg_price
AN8068571086	2012-11	69.69950
AN8068571086	2012-12	70.65000
AN8068571086	2013-01	75.06095
AN8068571086	2013-02	78.60211

#### Part 3 Input:

msci\_data from MSCIData\_2014\_to\_2018\_2021\_to\_2022.csv

ASOF_DATE	ISSUER_ISIN	INDUSTRY_ADJUSTED_SCORE	WEIGHTED_AVERAGE_SCORE	IVA_INDUSTRY	ENVIRONMENTAL_PILLAR_SCORE	GOVERNANCE_PILLAR_SCORE	SOCIAL_PILLAR_SCORE	year_month
2014-01-01	US0236081024	2.50	3.9	Utilities	3.2	8.5	6.9	2014-01
2014-01-01	US09247X1019	8.00	6.5	Asset Management	5.0	3.0	6.8	2014-01
2014-01-01	US7433151039	4.80	4.5	Property & Casualty Insurance	5.2	6.6	3.1	2014-01
2014-01-01	US1101221083	5.10	4.2	Pharmaceuticals	4.8	3.8	4.4	2014-01
2014-01-01	US2479162081	7.30	5.4	Oil & Gas Exploration & Production	5.0	7.1	4.3	2014-01

#### Part 4:

Combine average price and ESG score

ISIN_CD	year_month <sup>‡</sup>	avg_price <sup>‡</sup>	avg_environmental_score	avg_social_score	avg_governance_score
AN8068571086	2014-12	84.97500	6.4	5.3	5.7
AN8068571086	2015-01	81.43250	6.4	5.3	5.7
AN8068571086	2015-02	85.89789	6.4	5.3	5.0
AN8068571086	2015-03	82.61773	6.4	5.3	5.0

# Part 5 Output:

Calculated correlations per ESG score to price (correlations\_msci.csv)

ISIN_CD <sup>‡</sup>	environmental_score_correlation =	social_score_correlation +	governance_score_correlation +
AN8068571086	-0.136991869	-0.46768064	-0.442566207
BMG3223R1088	-0.777577153	0.76521218	-0.518748805

# GICS Sector Average Model

### Part 1 Input:

Price\_Data from PriceData\_2012\_2022.csv

DATE_DIM_ID *	BB_TICKER_CD	ISIN_CD	PBD_PRICE_AMT	year_month <sup>‡</sup>
2014-08-29	ZTS US	US98978V1035	35.44	2014-08
2014-09-04	ZTS US	US98978V1035	35.51	2014-09
2014-09-10	ZTS US	US98978V1035	36.24	2014-09
2014-09-12	ZTS US	US98978V1035	36.18	2014-09

### Part 2 Input:

MSCI\_Correlation\_Data from fixed\_10\_Years\_Correlation\_Table.csv

isinID	governancePillarScore	environmentalPillarScore <sup>‡</sup>	socialPillarScore <sup>‡</sup>
US0605051046	0.9213322	0.61607059	0.240115647
US4410601003	0.9083940	-0.79047968	-0.334921731
US37247D1063	0.8986127	0.90223321	-0.890752266
US46625H1005	0.8972508	0.49147501	0.041393895

#### Part 3 Input:

gics\_sector\_df from GICS\_Sectors.csv

isin_ID <sup>‡</sup>	sector_level_1
US8760301072	Consumer Discretionary
US00971T1016	Information Technology
US0010551028	Financials
US42250P1030	Real Estate

# Part 4 Output:

joined together MSCI to Price correlation with GICS Sector

isin_ID <sup>‡</sup>	environmental_correlation	social_correlation ‡	goverance_correlation	sector_level_1
AN8068571086	-0.810610746	-0.68733490	-0.680363829	Energy
BMG3223R1088	-0.024003061	0.06896253	-0.567242870	Financials
BMG491BT1088	-0.585447206	0.04789883	0.011421061	Financials
BMG667211046	-0.749440974	-0.71909902	-0.916910533	Consumer Discretionary
CH0044328745	-0.853012567	0.39099629	-0.404151996	Financials

### Part 5 Output:

entire\_joined\_sectors\_avg (calculates the avg correlation by GICS Sector)

sector_level_1	environmental_correlation	social_correlation +	goverance_correlation
Communication Services	0.30697120	0.2397773	-0.30882609
Consumer Discretionary	-0.10906882	0.2202777	-0.04968211
Consumer Staples	0.21456964	0.2618437	-0.01998541
Energy	0.22552648	-0.0272824	0.05848824
Financials	0.09299335	0.1308040	-0.03364343
Health Care	0.13422930	0.1944427	0.02258770
Industrials	0.07753924	0.2086654	-0.03177065
Information Technology	-0.02986816	0.1659122	-0.16472368
Materials	0.11714548	-0.1056803	0.03065386
Real Estate	0.27406580	-0.1492297	-0.31957017
Utilities	0.50321401	0.2702451	-0.13130124

# CQ Score to MSCI (ESG) Correlations Model

Part 1 Input:					
df from cq.csv					
	date =	ticker <sup>‡</sup>	isin_ID	price ‡	cq_score ‡
	20160114	ABT US	US0028241000	41.1	8.6
	20160104	ABT US	US0028241000	42.93	8.6
	20160223	ABT US	US0028241000	38.39	8.6
	20160209	ABT US	US0028241000	37	8.6

Part 2:										
Separate Mor	eparate Month and Year for CQ Score (cq_agg)									
	ticker	Month	Year ‡	Month_Year_Together	price <sup>‡</sup>	cq_score ‡				
	AA US	01	2015	01/2015	15.662500	6.760000				
	AA US	01	2016	01/2016	7.532105	5.636842				
	AA US	02	2015	02/2015	15.829474	6.900000				
	AA US	02	2016	02/2016	8.095000	5.500000				
	AA US	03	2015	03/2015	13.537273	6.900000				

Part 3 Input:											
Read MSCI data from ESG_Price_Date_Data.csv											
ticker	isin_ID ‡	price ‡	month ‡	year ‡	price.1 <sup>‡</sup>	environmental_pillar_score	social_pillar_score	governance_pillar_score			
ZTS US	US98978V1035	35.17167	8	2014	35.17167	4.8	7.7	3.0			
ZTS US	US98978V1035	36.33333	9	2014	36.33333	4.8	7.7	3.0			

Part 4	art 4:												
Comb	ine M	SCI S	Score and CQ	Score of	data fran	nes to c	reate df4						
ticker.x	Month	Year	Month_Year_Together	cq_score ÷	isin_ID	price.1	environmental_pillar_score	social_pillar_score	governance_pillar_score				
AA US	01	2015	01/2015	6.760000	US0138171014	15.662500	4.0	6.6	5.1				
AA US	02	2016	02/2016	5.500000	US0138171014	8.095000	4.8	5.8	5.0				
AA US	05	2015	05/2015	6.600000	US0138171014	13.372000	4.6	7.0	5.3				
AA US	05	2016	05/2016	5.500000	US0138171014	9.578571	4.8	5.8	5.5				
AA US	07	2016	07/2016	5.500000	US0138171014	10.343000	4.8	5.8	5.				
	08	2016	08/2016	5.500000	US0138171014	10.310000	4.8	5.8	4.5				
AA US	00												

### Part 5 Output:

Correlate each MSCI Score to CQ Score (CQ\_Score\_to\_ESG\_MSCI\_Correlation)

ticker.x <sup>‡</sup>	governancePillarScore <sup>‡</sup>	environmentalPillarScore <sup>‡</sup>	socialPillarScore <sup>‡</sup>
AA US	0.19055302	-0.853243143	0.942696582
AAP US	-0.64357626	-0.477418234	-0.394423797
AAPL US	-0.46356127	-0.530617373	0.163287973
ABBV US	-0.00473079	-0.441724451	-0.509702357

# CQ Score to Price Correlations Model

The model follows the same steps as the CQ Score to MSCI (ESG) Correlations Model outlined above. The difference is the output as the correlation coefficient is CQ to Price versus MSCI.

Final Output:	nal Output:									
Correlate each Price to CQ Score (correlations)										
	ticker <sup>‡</sup>	correlation_coefficient $^{\hat{ au}}$								
	AA US	0.427763256								
	AAL US	0.157085577								
	AAP US	0.114233007								
	AAPL US	0.107405482								
	ABBV US	-0.587949779								
	ABMD US	0.036775511								

# TruCost to Price Correlations Model

#### Part 1 Input: stock\_price\_data from PriceData\_2012\_2022.csv DATE\_DIM\_ID BB\_TICKER\_CD \* ISIN\_CD year\_month 2014-08-29 ZTS US US98978V1035 35.44 2014-08 2014-09-04 ZTS US US98978V1035 35.51 2014-09 2014-09-10 ZTS US US98978V1035 36.24 2014-09

Part 2:							
Calculate mor	nthly average	TruCos	st and F	Price (n	nonthly_avg_t	tru_cost) & (m	nonthly_avg_price
ISIN_CD	year_month +	avg_pric	е ‡				
AN8068571086	2012-11	69.6	9950				
AN8068571086	2012-12	70.6	5000				
AN8068571086	2013-01	75.0	6095				
ISIN <sup>‡</sup>	GICSSECTORNA	ME ‡	year_m	onth <sup>‡</sup>	avg_scope1	avg_scope2	avg_first_tier ‡
AN8068571086	Energy		2017-0	4	3.962113e+01	16.3495419	190.182072
AN8068571086	Energy		2020-1	0	4.582159e+01	18.6291370	199.882287
AN8068571086	Energy	0,		8	4.099200e+01	25.1830000	201.182000

Part 3:										
Combine and clean TruCost data (combined_data)										
	ISIN_CD	year_month	avg_price	avg_scope1 ‡	avg_scope2	avg_first_tier				
	AN8068571086	2017-04	76.78789	39.6211296	16.349542	190.182072				
	AN8068571086	2017-05	71.28864	39.6211296	16.349542	190.182072				
	AN8068571086	2017-06	67.57864	39.6211296	16.349542	190.182072				

## Part 4:

# Calculate correlation coefficient between Price and TruCost (correlations\_with\_ticker)

BB_TICKER_CD *	carbon_intensity_direct_and_first_tier_indirect_correlation	carbon_emissions_scope_1_correlation	carbon_emissions_scope_2_correlation +
ZTS US	0.664947898	-0.877261635	0.774581142
ZMH US	0.283502686	0.175567421	0.421709955
ZION US	-0.219434326	-0.208181034	-0.186459612
ZBRA US	0.139517132	0.139517132	0.139517132
ZBH US	0.283502686	0.175567421	0.421709955
	ZTS US ZMH US ZION US ZBRA US	ZTS US 0.664947898  ZMH US 0.283502686  ZION US -0.219434326  ZBRA US 0.139517132	ZTS US     0.664947898     -0.877261635       ZMH US     0.283502686     0.175567421       ZION US     -0.219434326     -0.208181034       ZBRA US     0.139517132     0.139517132