Responsible investing: The ESG-efficient frontier

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Outline

- Introduction
- Portfolio choice with ESG: the ESG-efficient frontier
- Equilibrium asset pricing with ESG
- Empirical results
- Conclusion

1. Introduction-- Motivation

- Investors have little guidance in how to incorporate ESG in portfolio choice
- Opinions differ dramatically about whether ESG will help or hurt their performance.
- To reconcile these opposing views, we develop a theory that illuminates both the potential costs and benefits of ESG-based investing.

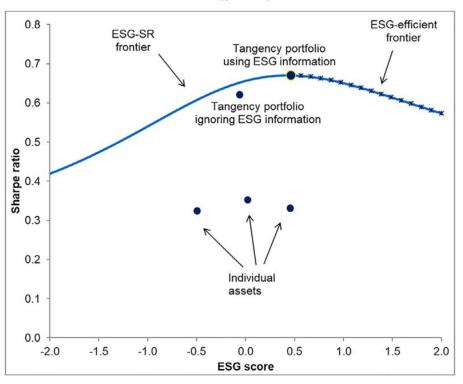
1. Introduction-- Contribution

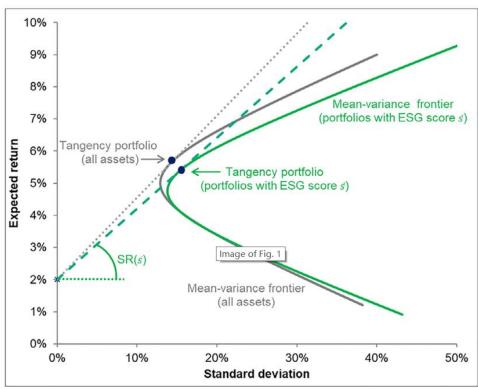
- We contribute to the literature on ESG both theoretically and empirically.
- Theoretically, We explicitly model many assets characterized by ESG scores in addition to the standard risk-return characteristics.
- Empirically, our research bridges the gap between papers arguing that ESG hurts performance and those arriving at the opposite conclusion.

1. Introduction-- Content

Panel B: Mean-variance frontiers for all assets and portfolios with certain ESG score

Panel A: ESG-efficient frontier





- n risky assets and a risk-free security.
- The risk free return is r^f
- The risky assets have excess returns collected in the vector of random variables denoted by $r = (r^1, \dots r^n)'$.
- The assets have an ESG scores given by $s = (s^1, ... s^n)'$.
- Type-U investors: $\mu = E(r)$, $\Sigma = var(r)$.
- Type-A investors: $\mu = E(r|s)$, $\Sigma = var(r|s)$.
- Type-M investors: $\mu = E(r|s)$, $\Sigma = var(r|s)$ and also have **preferences** for high ESG scores

Investor M starts with a wealth W and a portfolio of risky assets

$$x = (x^{1}, ..., x^{n})'$$

$$\widehat{W} = W\left(1 + r^{f} + x'r\right).$$

$$U = E\left(\widehat{W}|s\right) - \frac{\bar{\gamma}}{2}Var\left(\widehat{W}|s\right) + Wf\left(\bar{s}\right). \qquad \bar{s} = \frac{x's}{x'1}$$

$$U = W\left(1 + r^{f} + x'\mu\right) - \frac{\bar{\gamma}}{2}W^{2}x'\Sigma x + Wf\left(\frac{x's}{x'1}\right)$$

$$= W\left(1 + r^{f} + x'\mu - \frac{\gamma}{2}x'\Sigma x + f\left(\frac{x's}{x'1}\right)\right), \qquad \gamma = \bar{\gamma}W$$

$$\max_{x \in X} \left(x'\mu - \frac{\gamma}{2}x'\Sigma x + f\left(\frac{x's}{x'1}\right)\right), \qquad X = \{x \in \mathbb{R}^{n} | x'1 > 0\},$$

Solution: ESG-SR frontier

$$SR(\bar{s}) = \max_{\substack{\chi \in X \\ \text{s.t. } \bar{s} = \frac{\chi's}{\chi'1}}} \left(\frac{\chi'\mu}{\sqrt{\chi'\Sigma\chi}}\right) = \max_{\substack{\chi \\ \text{s.t. } \chi'1 = 1 \\ \text{and } \chi's = \bar{s}}} \left(\frac{\chi'\mu}{\sqrt{\chi'\Sigma\chi}}\right)$$

$$\max_{\bar{s}} \left[\max_{\sigma} \left\{ \max_{\substack{x \in X \\ \text{s.t. } \bar{s} = \frac{x's}{x'1} \\ \sigma^2 = x' \sum x}} \left(x' \mu - \frac{\gamma}{2} \sigma^2 + f(\bar{s}) \right) \right\} \right]. \quad (6)$$

$$\max_{\bar{s}} \left[\max_{\sigma} \left\{ SR(\bar{s})\sigma - \frac{\gamma}{2} \ \sigma^2 + f(\bar{s}) \right\} \right]. \tag{7}$$

对公式(7)求一阶导数 $\sigma = SR(s)/\gamma$

> Solution: ESG-SR frontier

Proposition 1 (ESG-SR trade-off). The investor should choose her average ESG score \bar{s} to maximize the following function of the squared Sharpe ratio and the ESG preference function f

$$\max_{\bar{s}} \left[(SR(\bar{s}))^2 + 2\gamma f(\bar{s}) \right]. \tag{8}$$

Proposition 2 (ESG-SR frontier). The maximum Sharpe ratio, $SR(\bar{s})$, that can be achieved with an ESG score of \bar{s} is

$$SR(\bar{s}) = \sqrt{c_{\mu\mu} - \frac{\left(c_{s\mu} - \bar{s}c_{1\mu}\right)^2}{c_{ss} - 2\bar{s}c_{1s} + \bar{s}^2c_{11}}}.$$
 (9)

$$SR(s^*) = \sqrt{c_{\mu\mu}}, \qquad s^* = c_{s\mu}/c_{1\mu}. \qquad c_{ab} = a' \Sigma^{-1} b$$

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> Solution: ESG-SR frontier

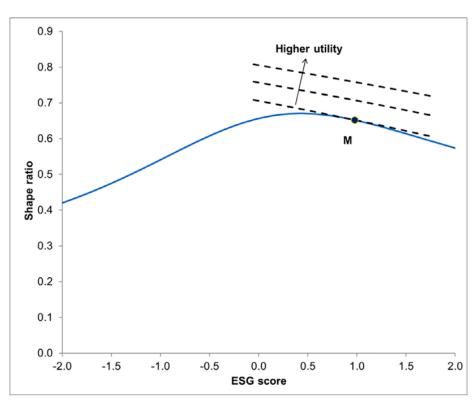
Proposition 3 (four-fund separation). Given an average ESG score \bar{s} , the optimal portfolio is

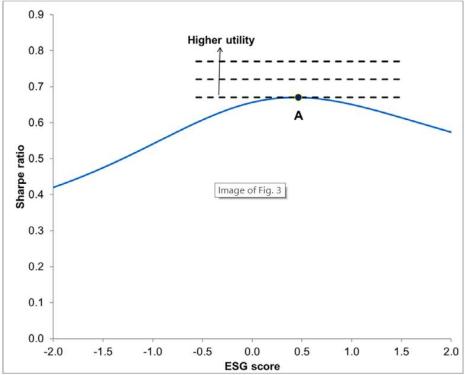
$$x = \frac{1}{\gamma} \Sigma^{-1} (\mu + \pi (s - 1\bar{s}))$$
 (10)

The optimal portfolio is therefore a combination of the risk-free asset, the tangency portfolio, $\Sigma^{-1}\mu$, the minimum-variance portfolio, $\Sigma^{-1}1$, and the ESG-tangency portfolio, $\Sigma^{-1}s$.

Example: how investors choose portfolios using the ESG-SR frontier

Panel A: Indifference curves for an ESG-motivated investor (type-M) Panel B: Indifference curves for an ESG-aware investor (type-A)





Generalized ESG preferences

Proposition 4 (ESG-SR frontier with screens). The conclusion of Proposition 1 continues to hold for any cone-shaped X.

$$X = \{ x \in \mathbb{R}^n | x'1 > 0, \forall i \ x^i = 0 \text{ if } s^i < s^* \}$$
$$X = \{ x \in \mathbb{R}^n_+ | \forall i \ x^i = 0 \text{ if } s^i < s^* \}$$

Proposition 5 (generalized ESG-SR frontier). If the investor has generalized ESG preferences e(x,s), then the investor's problem is

$$\max_{\bar{e}} \left[\frac{(SR(\bar{e}))^2}{2\gamma} + \bar{e} \right], \tag{12}$$

Example: ESG scores: (0.1, 0.8, 0.9) VS (0.6, 0.6, 0.6)

$$e(x,s) = e_1 \frac{x's}{x'1} - e_2 \frac{x'\operatorname{diag}(\frac{1}{s_1}, \dots, \frac{1}{s_n})x}{(x'1)^2}$$

3. Equilibrium asset pricing with ESG

> ESG-adjusted CAPM

Proposition 6. If all investors are ESG-unaware, i.e., of type-U

 $(W_A = W_M = 0)$, then any security i has steady-state equilibrium price

$$p^{i} = \frac{\hat{\mu}^{i} - \frac{\gamma}{W} \operatorname{cov}(v^{i}, v^{m})}{r^{f}}.$$

Unconditional expected excess return obeys the standard unconditional CAPM:

$$E(r_t^i) = \beta^i E(r_t^m),$$

But conditional expected returns are given by

$$E(r_t^i|s) = \beta^i E(r_t^m) + \lambda \frac{s^i - s^m}{p^i}.$$

3. Equilibrium asset pricing with ESG

ESG-adjusted CAPM

$$p^i = rac{\hat{\mu}^i - rac{\gamma}{W} \, \operatorname{cov}ig(v^i, v^mig)}{r^f}.$$

Proposition 7 (ESG-CAPM). If all investors are ESG-motivated of type-M $(W_U = W_A = 0)$, then any security i has equilibrium price

$$p^{i} = \frac{\hat{\mu}^{i} + \lambda \left(s^{i} - s^{m}\right) - \frac{\gamma}{W} \operatorname{cov}(v^{i}, v^{m}|s)}{r^{f} - \pi \left(s^{i} - s^{m}\right)},$$
(20)

The equilibrium conditional expected excess return is given by

$$E(r_t^i|s) = \bar{\beta}^i E(r_t^m|s) - \pi \left(s^i - s^m\right). \tag{21}$$

If all investors are ESG-aware of type-A ($W_U = W_M = 0$), the same conclusions hold with $\pi = 0$.

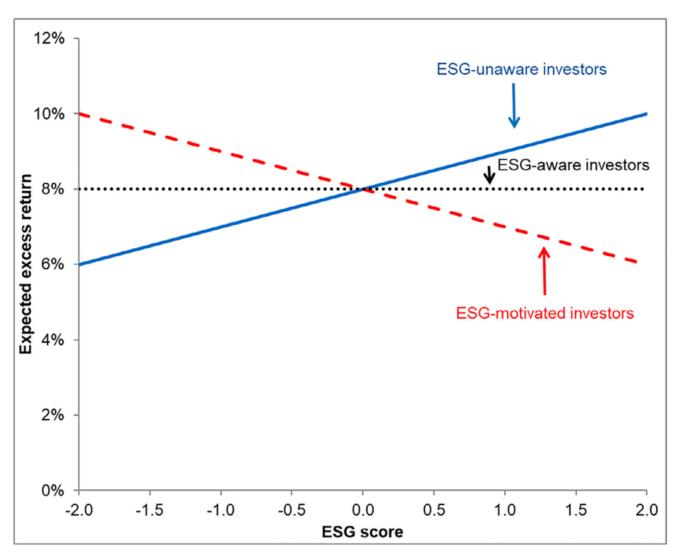


Fig. 2. Environmental, social, and governance-adjusted capital asset pricing model (ESG-CAPM).

> Data

- Stocks in the Standard & Poor's (S&P) 500 index
- E: low carbon intensity, 2009.01~ 2019.03
 (Busch et al., 2018)
- S: non-sin stock indicator. zero for sin stocks and the value of one otherwise, 1963.01~2019.03 (Hong and Kacperczyk, 2009)
- **G**: low accruals. 1963.01~2019.03 (Sloan,1996))
- overall **ESG**: MSCI ESG scores(0~10), 2007.01~ 2019.03
- XpressFeed database : stock returns and market values
- Compustat database : firm fundamentals,

Empirical ESG-SR frontier

To compute the annualized expected return of any stock i in any month t,
 U investors use

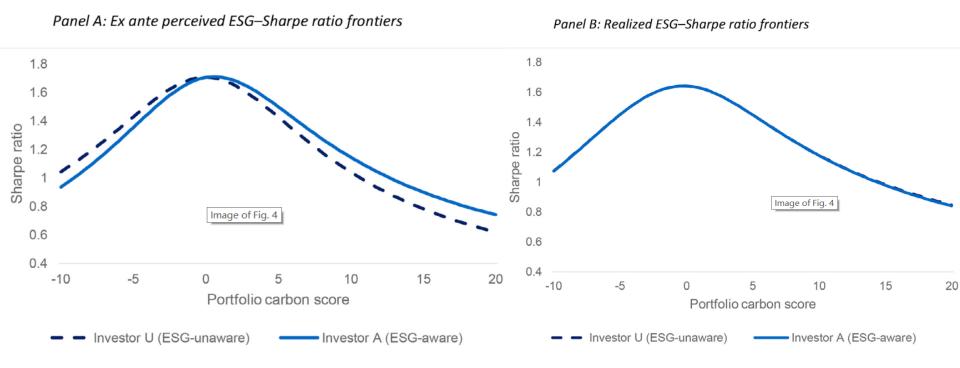
$$E_t^{\mathsf{U}}(r_{i,t+1}) = \overline{MKT}_t + bm_{i,t} \ \overline{BM}_t, \tag{23}$$

A and M investors

$$E_t^{A}(r_{i,t+1}) = \overline{MKT}_t + bm_{i,t} \ \overline{BM}_t + s_{i,t} \ \overline{ESG}_t, \tag{24}$$

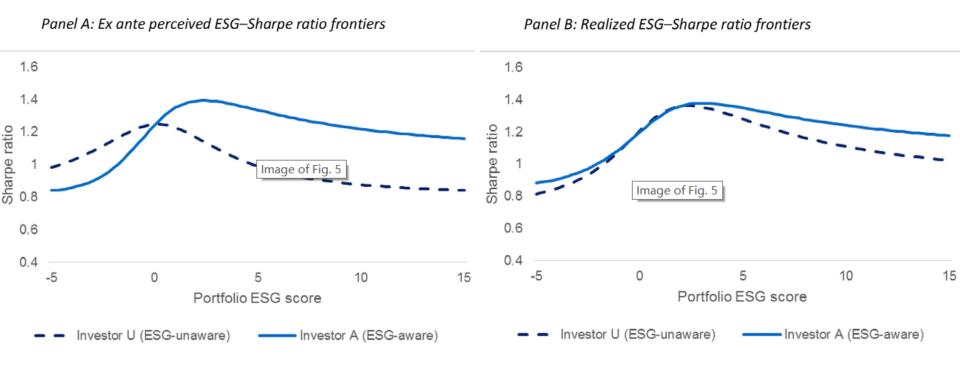
- $\overline{ESG_t}$ is the return premium of the ESG factor, the ESG score $s_{i,t}$ is computed as the cross-sectional z-score of the raw ESG metric.
- To compute risk, Barra's US Equity risk model (Barra USE3L model)

Empirical ESG-SR frontier



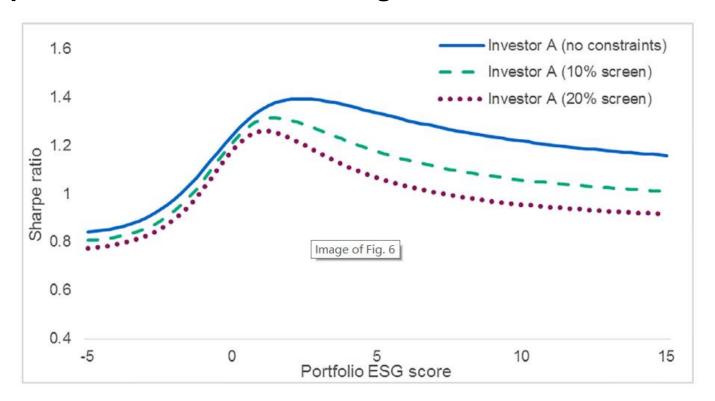
• The environmental proxy we use here is **not very helpful** in explaining average returns.

Empirical ESG-SR frontier



- Empirical ESG–efficient frontier using accruals as a proxy for G.
- The benefit of using G information is 11% higher than the realized SR of the ESG-unaware investor

Impact of restrictions: screening out the worst ESG stocks



- Empirical ESG–efficient frontier using accruals as a proxy for G.
- Constraints reduce a portfolio's expected performance.

- RNOA: return on net operating assets
- Gross profit over assets

Does ESG predict future fundamentals?

Panel A: Predicting RNOA								
Dependent variable	$RNOA\ (t+12)$							
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
E (low CO2)	0.006*** (4.91)	0.006*** (7.34)						
S (non-sin)	(33.2)	(100.07)	-0.008* (-1.94)	-0.006*** (-2.88)				
G (low accruals)				<u>, , , , , , , , , , , , , , , , , , , </u>	0.208*** (23.26)	0.193*** (28.64)		
ESG (MSCI)							0.0001 (0.15)	0.0001 (0.24)
Beta	-0.068*** (-17.90)	-0.068*** (-10.24)	-0.064*** (-33.77)	-0.067*** (-20.69)	-0.060*** (-31.79)	-0.062*** (-19.43)	-0.052*** (-11.62)	-0.040*** (-4.40)
Ln market cap	0.011*** (12.45)	0.011*** (23.91)	0.015***	0.015*** (26.55)	0.014*** (30.14)	0.014*** (26.85)	0.008***	0.006*** (4.89)
Ln(P/B)	0.014*** (6.72)	0.015*** (6.98)	0.027*** (22.59)	0.028*** (22.01)	0.028*** (23.73)	0.028*** (22.11)	0.026*** (9.27)	0.038*** (11.94)
RNOA(t)	0.763*** (88.59)	0.765*** (97.48)	0.710*** (167.53)	0.707*** (118.95)	0.725*** (169.65)	0.720*** (128.80)	0.756*** (63.53)	0.734*** (61.25)
Constant	0.020*** (2.78)	0.021** (2.32)	-0.005 (-0.95)	0.003 (0.47)	-0.019*** (-6.59)	-0.009 (-1.56)	0.002 (0.19)	0.001 (0.06)
Number of observations <i>R</i> -squared	239,440 0.708	239,440 0.712	1374,620 0.631	1374,620 0.631	1354,499 0.636	1354,499 0.635	116,130 0.723	116,130 0.727
Estimation method	Pooled	FM	Pooled	FM	Pooled	FM	Pooled	FM

There is strong evidence that accruals correlate with future profitability

- institutional ownership,
- trading activity: logarithm of the number of trades
- signed order flow :dollar buy volume over total dollar volume

Does ESG predict investor demand?

Panel A: Predicting institutional ownership								
Dependent variable	Institutional holdings $(t + 3)$							
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
E (low CO2)	2.206***	2.284***						
	(3.37)	(14.65)						
S (non-sin)			6.128**	7.037***				
			(2.43)	(11.50)				
G (low accruals)					1.060	3.208***		
					(0.74)	(2.98)		
ESG (MSCI)							0.343**	0.420***
							(2.55)	(6.98)
Beta	5.774***	5.912***	5.698***	6.905***	1.610***	3.038***	6.371***	5.512***
	(8.50)	(21.96)	(14.13)	(20.76)	(3.37)	(11.91)	(7.05)	(11.27)
Ln market cap	10.079***	10.057***	9.662***	9.691***	9.599***	9.650***	0.846***	-1.265**
	(50.48)	(108.99)	(62.30)	(64.95)	(53.67)	(85.18)	(3.32)	(-2.67)
Ln(P/B)	-0.321	-0.354***	-1.759***	-1.264***	-2.282***	-1.931***	1.136***	1.642***
	(-1.20)	(-5.08)	(-11.05)	(-8.39)	(-13.90)	(-13.83)	(3.86)	(9.22)
Constant	-10.649***	-10.400***	-17.176***	-19.342***	-3.402***	-5.076***	62.372***	82.049***
	(-6.77)	(-17.28)	(-6.40)	(-18.11)	(-3.00)	(-9.55)	(24.56)	(18.45)
Number of observations	378,623	378,623	962,867	962,867	737,865	737,865	180,326	180,326
R-squared	0.454	0.450	0.470	0.424	0.475	0.422	0.033	0.083
Estimation method	Pooled	FM	Pooled	FM	Pooled	FM	Pooled	FM

 The results are perhaps most intuitive for accruals, where both the number of trades and the fraction of buys increase when this ESG proxy improves

• Firm's valuation ratio: the logarithm of price-to-book

Does ESG predict valuation and future returns?

Dependent variable	Ln(P/B)						
	(1)	(2)	(3)	(4)			
E (low CO2)	0.086*** (7.25)						
S (non-sin)		0.020 (0.30)					
G (low accruals)			-0.470*** (-11.59)				
ESG (MSCI)				0.058*** (8.25)			
Beta	-0.449*** (-16.39)	0.402*** (28.48)	0.338*** (21.13)	-0.348*** (-8.56)			
Constant	1.391*** (38.32)	0.366*** (5.48)	0.514*** (27.37)	1.245*** (21.81)			
Number of observations R-squared Estimation method	427,857 0.050 Pooled	2120,679 0.073 Pooled	1708,222 0.077 Pooled	203,502 0.046 Pooled			

 Shows how the ESG proxies correlate with the logarithm of the price-tobook ratio.

Does ESG predict valuation and future returns?

	Е	S	G	ESG
	(low CO ₂)	(non-sin)	(low accruals)	(MSCI)
Panel A: Equal-weighted returns				
Average excess return	5.15%	0.50%	7.84%***	0.38%
	(1.59)	(0.35)	(4.41)	(0.28)
CAPM alpha	7.02%**	-0.42%	7.87%***	1.29%
	(2.09)	(-0.30)	(4.39)	(1.00)
Three-factor (FF) alpha	5.03%	0.06%	7.30%***	0.74%
	(1.63)	(0.05)	(4.03)	(0.60)
Five-factor (FF) alpha	5.98%*	1.28%	8.85%***	0.28%
	(1.92)	(0.94)	(4.91)	(0.22)
Six-factor (FF + Mom) alpha	5.12%*	1.03%	8.71%***	0.27%
	(1.73)	(0.74)	(4.76)	(0.22)

The portfolio based on G has highly significant returns.

6. Conclusion

- We show that an investor optimally chooses a portfolio on the ESG efficient frontier both theoretically and empirically.
- We test the theory's equilibrium predictions using four ESG proxies, providing a rationale for why certain ESG measures predict returns differently.

Reflection

The model can be used to invest in assets with specific themes,
 such as scientific and technological innovation capability.