Weekly Meeting Agenda



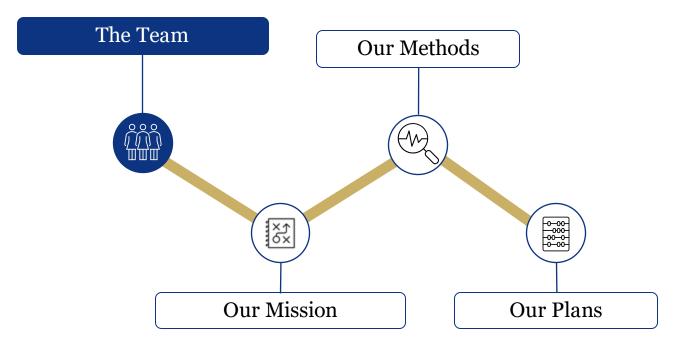
01

Specialized Teams Pitches

Our Specialized Teams Will Present Themselves and Their Projects For This Semester









The Team



- Alexandre Martens
- Senior Risk Analyst
- Lead developer



- Max Malata
- Junior Analyst
- Portfolio Optimization



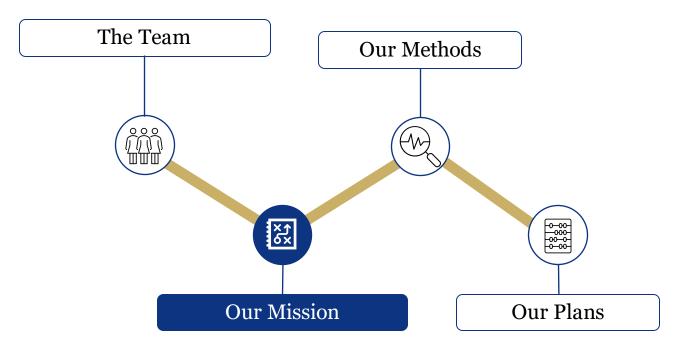


- Constantin Hinz
- Junior Analyst
- Geographic and Currency Risk
- Victor den Ouden
- Junior Analyst
- Risk Simulation



- Epifanios Evangelou
- Junior Analyst
- Value-at-Risk

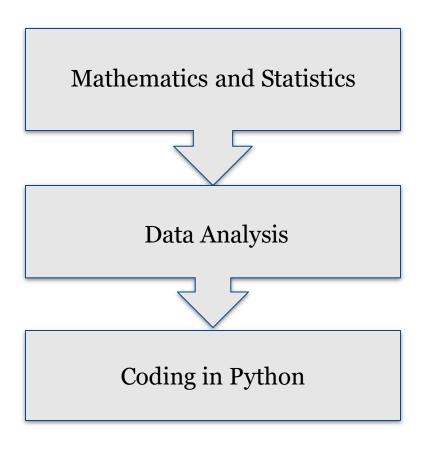






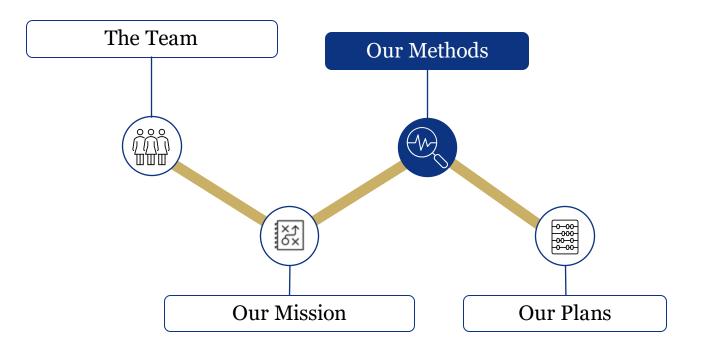
About the Risk Team

What do we do?



Why do we do it?

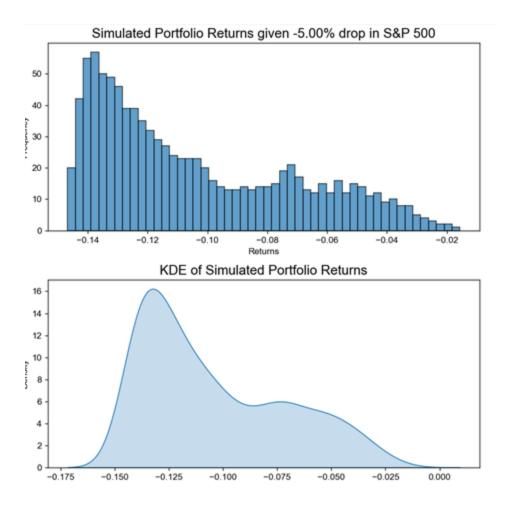
- Identify risk profile
- Protect fund against large losses
- Reduce concentration of risk
- Locate and evaluate potential additions





Copula Functions

Calculating Effects



Application

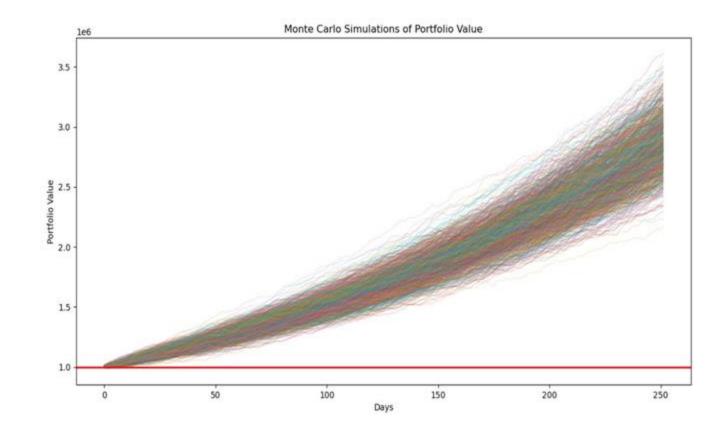
- Financial markets are not always normally distributed
- Multiple univariate distributions → one multivariate distribution
- Normalizes marginal distributions
- Enables comparison

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Monte Carlo Simulation

Estimating Outcomes

- Shows the spectrum of probable portfolio outcomes
- Repeatedly assigns random values to variables
- Helps explain impact of risk and uncertainty on a portfolio
- Assumes perfectly efficient market





Portfolio Optimisation

Finding the optimal allocation of assets

What is it all about?

- Finding the optimal allocation of assets for a specific goal
 - min risk
 - max return
 - max sharpe ratio
- Finding and balancing assets with low correlation to achieve diversification
- Building on the work of analysts and investors

Challenges and Factors to consider:

- Minimum and Maximum weights
 - per stocks
 - per industry
- Transaction costs
 - costs associated with rebalancing
- Maintaining high diversification *avoid over concentrated weights*
- Finding the best and most accurate risk measure

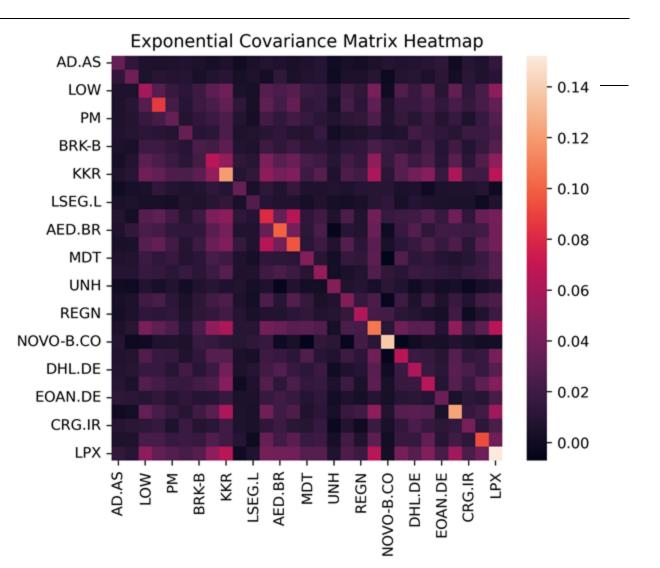
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Portfolio Optimisation

Finding the optimal allocation of assets

| | MV | MAD | MSV | FLPM | SLPM | CVaR | EVaR | WR | MDD | ADD | CDaR | UCI | EDaR |
|-----------|-----|------|------|-------|-------|------|------|-----|------|------|------|-----|------|
| AD.AS | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 | 0,03 | 0,06 | 0,01 | 0,1 | 0,04 |
| HEIA.AS | 0,1 | 0,06 | 0,07 | 0,055 | 0,063 | 0,03 | 0,06 | 0,1 | 0,1 | 0,07 | 0,1 | 0,1 | 0,1 |
| LOW | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,02 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| TGT | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| PM | 0,1 | 0,1 | 0,08 | 0,093 | 0,081 | 0,08 | 0,06 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 |
| ALV.DE | 0 | 0,06 | 0,04 | 0,076 | 0,045 | 0,04 | 0,1 | 0,1 | 0,01 | 0,05 | 0,01 | 0 | 0,01 |
| BRK-B | 0 | 0,02 | 0,03 | 0,046 | 0,049 | 0,06 | 0,03 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| SPGI | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,1 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| KKR | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| CME | 0,1 | 0,07 | 0,08 | 0,072 | 0,074 | 0,06 | 0,01 | 0 | 0,01 | 0,03 | 0,01 | 0 | 0,01 |
| LSEG.L | 0,1 | 0,1 | 0,1 | 0,088 | 0,1 | 0,1 | 0,08 | 0 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 |
| EQIX | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| AED.BR | 0 | 0,05 | 0,06 | 0,024 | 0,038 | 0,06 | 0,1 | 0,1 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| AMT | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| MDT | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| BAYN.DE | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,02 | 0,05 | 0,1 | 0,1 | 0,01 | 0,07 | 0 | 0,1 |
| UNH | 0,1 | 0,07 | 0,08 | 0,075 | 0,079 | 0,1 | 0,01 | 0 | 0,01 | 0,1 | 0,1 | 0,1 | 0,03 |
| LH | 0,1 | 0,05 | 0,05 | 0,044 | 0,047 | 0,04 | 0,08 | 0,1 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| REGN | 0 | 0,03 | 0,02 | 0,019 | 0,015 | 0,06 | 0,01 | 0 | 0,08 | 0,06 | 0,09 | 0,1 | 0,07 |
| IDXX | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| NOVO-B.CO | 0,1 | 0,07 | 0,07 | 0,1 | 0,095 | 0,05 | 0,08 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 |
| HCA | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| DHL.DE | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| CMI | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,1 | 0,1 | 0,1 | 0,1 | 0,1 |
| EOAN.DE | 0,1 | 0,06 | 0,06 | 0,048 | 0,055 | 0,03 | 0,09 | 0 | 0,01 | 0,06 | 0,01 | 0 | 0,01 |
| AMAT | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |
| CRG.IR | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,02 | 0,01 | 0 | 0,01 |
| APD | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,09 | 0,01 | 0,03 | 0 | 0,06 |
| LPX | 0 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0,01 | 0 | 0,01 | 0,01 | 0,01 | 0 | 0,01 |

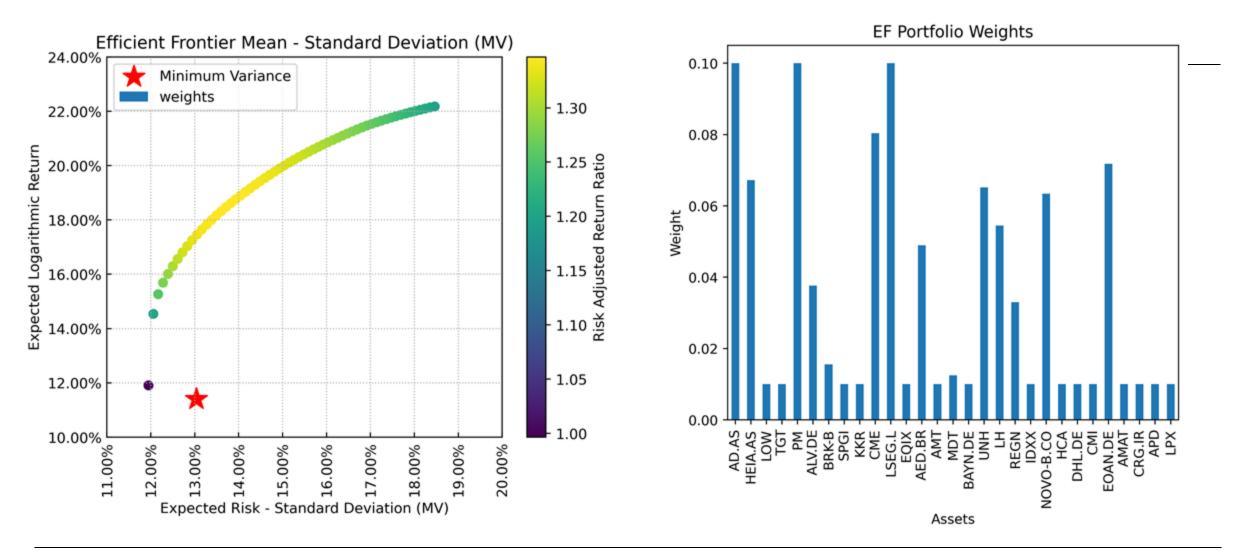




Efficient Frontier (EF) Portfolio



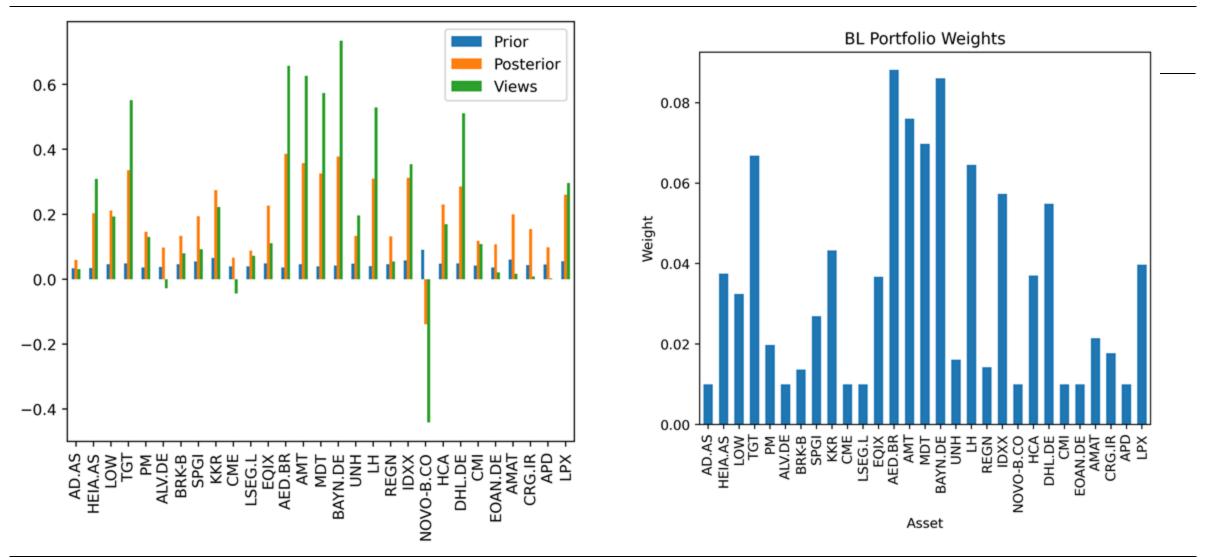
Methods based on classical portfolio theory



Black-Litterman (BL) Portfolio

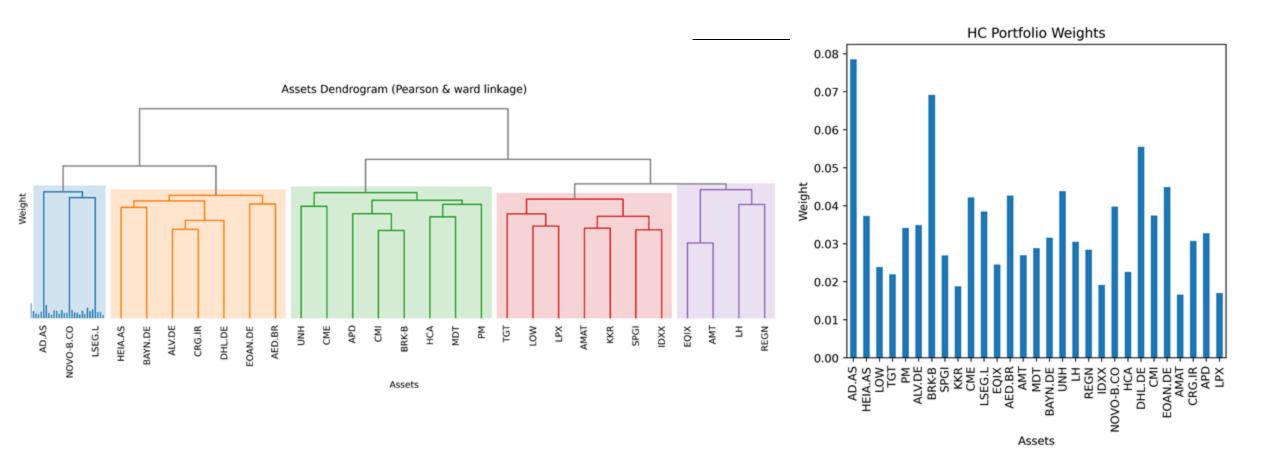


A widely-adopted model in the industry



Hierarchical Clustering (HC) Portfolio

An innovative and advanced model



Value-at-Risk

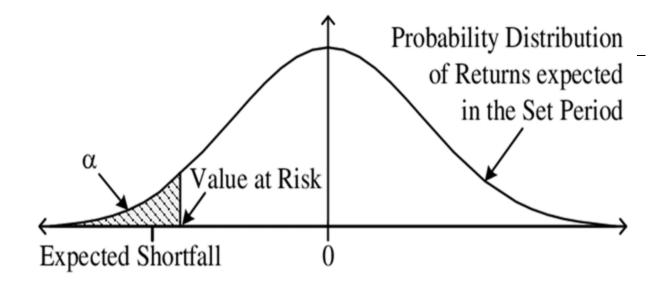
"What is the maximum amount I could lose over a specific period, with a certain level of confidence?"

VaR takes into account:

- 1. Volatility of investments,
- 2. Historical market data,
- 3. Probability distribution of potential losses.

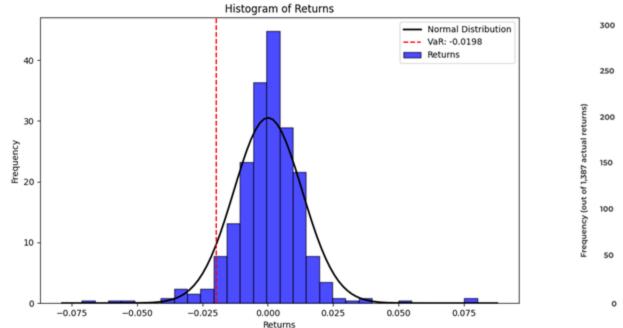
Types of VaR used in Sigma's Risk Team:

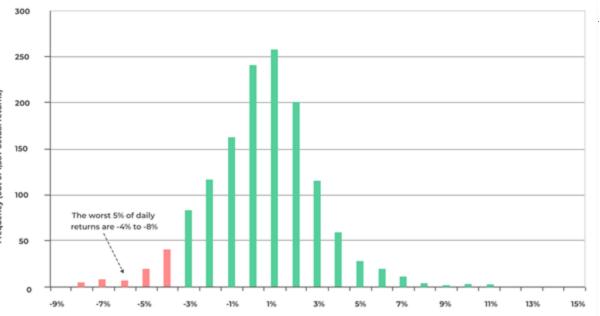
- 1. Historical VaR,
- 2. Gaussian (Parametric) VaR,
- 3. Marginal and Component VaR,
- 4. Incremental VaR (IvaR).
- 5. Expected Shortfall (or Conditional VaR, CVaR)



Historical and Gaussian VaR







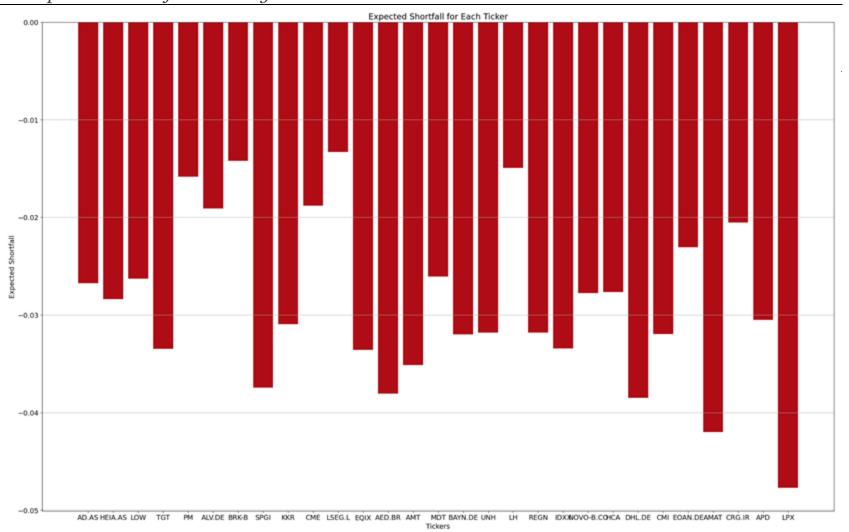
| | Gaussian (Parametric) VaR | Historical VaR | | | | |
|-------------|---|--|--|--|--|--|
| Definition | Parametric approach to calculate expected losses using approximately normally distributed returns | Non-parametric approach that uses historical data to estimate expected losses. | | | | |
| Methodology | Estimate mean and standard deviation of asset and use: z-score*std*mean | Calculate returns using historical prices and then select the nth percentile (e.g., 5th percentile for 95% confidence) as VaR, where n corresponds to the chosen confidence level. | | | | |
| Assumptions | Assumes normality of returns. | Relies on market data and can be sensitive to historical anomalies | | | | |

Expected Shortfall

Expected Shortfall (ES) is the negative of the expected value of the tail beyond the VaR

More comprehensive view of potential losses beyond VaR. Insight into the severity of extreme outcomes. Interpretation:

- 1. Average Loss Beyond VaR,
- 2. Severity of Extreme Losses,
- 3. Risk Management and Decision Making,
- 4. Comparison and Analysis.



Marginal and Component VaR

Your running security amongst the unpredictability of the floor.

Component VaR Analysis:

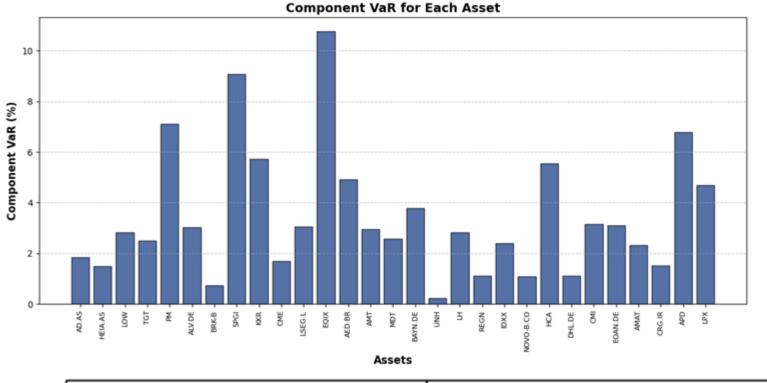
Decomposes VaR by portfolio components (e.g., assets, sectors).

•Risk Management Insights:

• Identifies key risk contributors, enhancing risk management.

•Diversification Benefits:

 Compares VaR of diversified vs. undiversified portfolios, evidencing Sigma's risk diversification success.



| Ticker | Component VaR (%) | | | |
|-------------------------|-------------------|--|--|--|
| Diversified Portfolio | 288.0777845537489 | | | |
| Undiversified Portfolio | 602.2107995872481 | | | |

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Incremental VaR

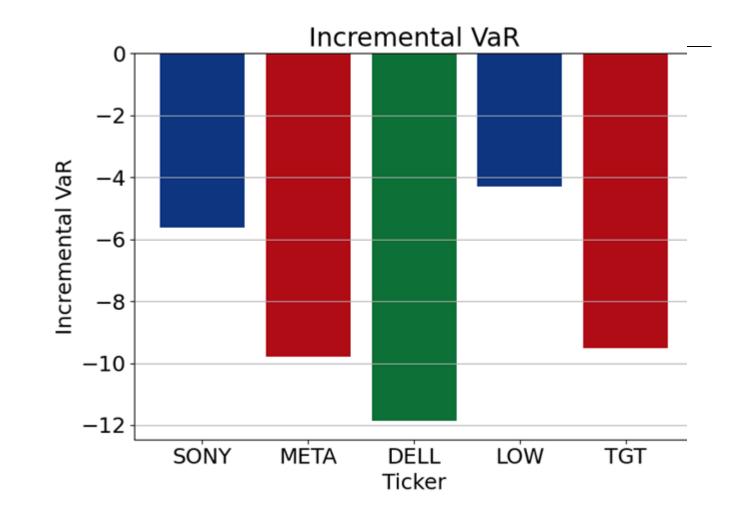
Somehow, we love to settle for the analytical solutions allowing us to "see a less attractive girl" – more attractive.

Incremental VaR:

-Reflects how the addition of a certain stock affects the value at risk of the portfolio,

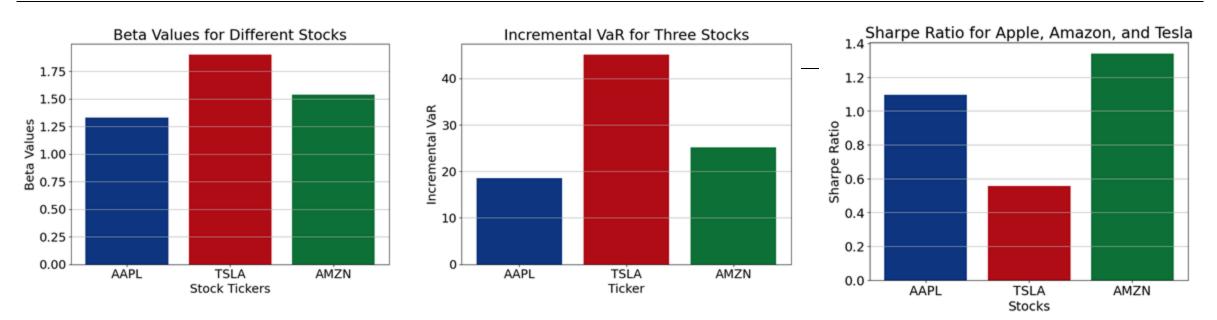
—Useful for deciding whether to add a certain asset to the portfolio,

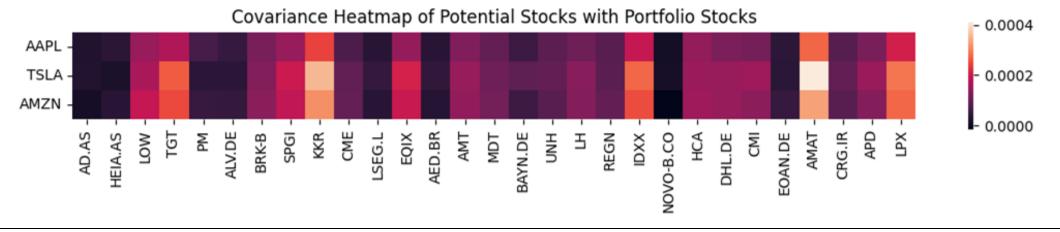
-Also useful when deciding which asset should be removed from the portfolio.



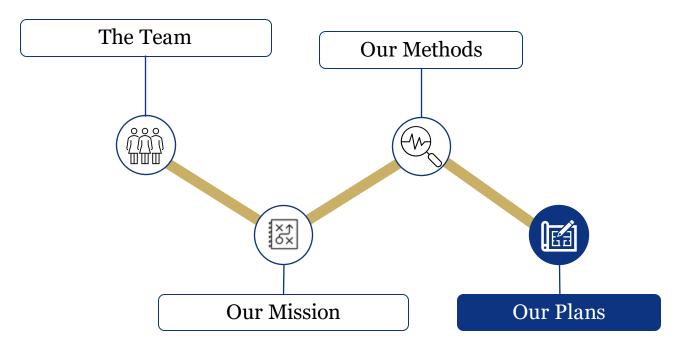
Pitch Evaluation

 $Compare \, stocks \, before \, further \, analysis$





Risk Team





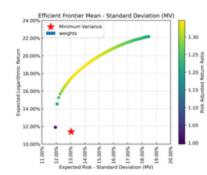
Future Prospects

We aim to add more metrics and summarize them in one easy to access document

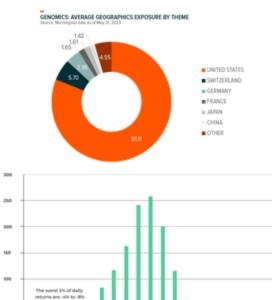


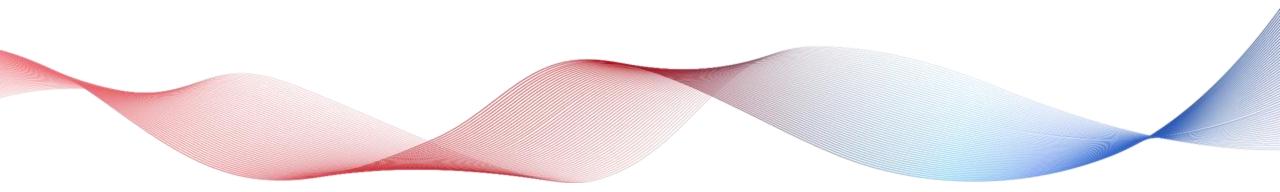
1. DCF model in Python

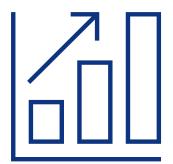
- \Rightarrow Faster stock analysis
- 2. Jump-Diffusion model
 - \Rightarrow More accurate predictions about future jumps
- 3. Refining VaR estimations
 - \Rightarrow More detailed risk assessment
- 4. Geographic and currency exposure





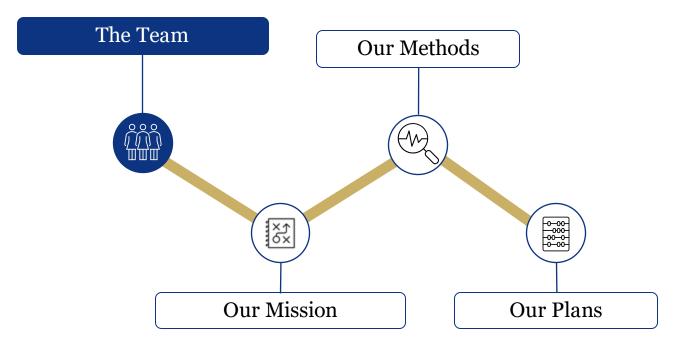






Technical Analysis







The Team



- Riccardo Supino
- Senior Analyst
- Technical Analysis



- Dario Saveriano
- Junior Analyst
- Technical Analysis



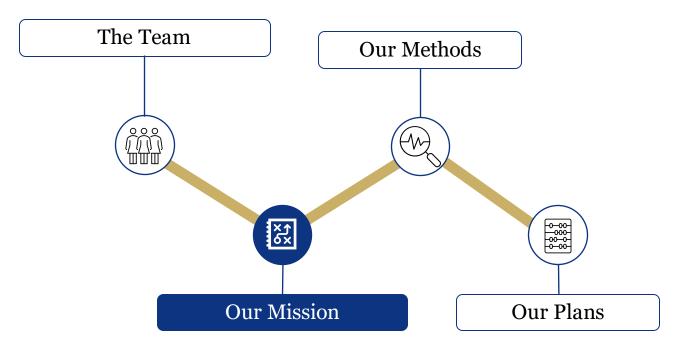
- Jordi Llompart
- Junior Analyst
- Technical Analysis



- Nicholas Friedrich
- Junior Analyst
- Quantitative & Machine Learning



- Maja Swietochowska
- Junior Analyst
- Statistics & R studio



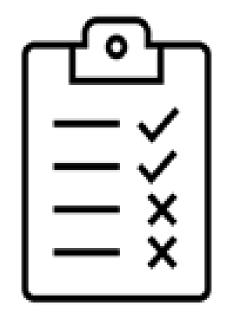


The Mission

Optimizing entry & exit positions

We analyze various indicators to paint a story

- We are **not** predicting the market
- Analysing market sentiment and technical indicators
- Work closely with industry teams & fund analyst team
- Incorporate Market sentiment & event news
- Constantly monitoring open positions in the fund portfolio
- Monthly reports



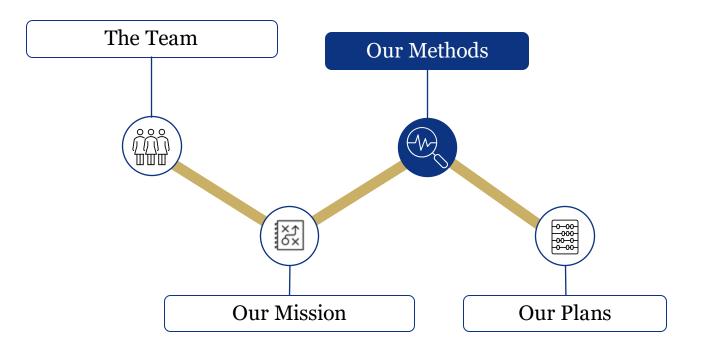
Why Technical Analysis works



Market Psychology

- Price Patterns that tend to repeat themselves.
 - Especially as more people trade using Technical Analysis.
 - Reflect underlying human psychology in trading decisions.
- Markets can hint at direction of next movement
- Mathematical indictors can portray characteristics of stock price movement









We use a variety of techniques in our analysis:

Non-Visual Indicators

Methodology

A Multifaceted Approach to Market Analysis

- Statistical Models
- Market Sentiment Analysis

Support & Resistance Channels

Σ

Analyzing the general direction

- **Support**: Price level at which a market or stock historically stops falling
- **Resistance**: Price level at which a market or stock historically struggles to surpass
- **Channels**: Represent the boundaries of a price range, showing trend direction
- Establish breakout points in the channels
- They can be influenced by psychological and historical factors



Moving Average Convergence/Divergence (MACD) Σ

A trend-following momentum indicator used to identify changes in direction of a trend

- MACD line
- Signal line
- MACD histogram
- A crossover of the MACD line above/below the signal line is often considered a buy/Sell signal.
- Divergence can be a warning sign that a trend is about to reverse.



Relative Strength Indicator (RSI)



A momentum oscillator measuring price movement speed and overbought/oversold conditions

- Identifying reversal points.
- Confirming trend strength.
- Spotting divergences between RSI and prices
- Divergence signals weakening trend.
- Steep/persistent RSI values indicate trend strength.



26.02.2024

Directional Movement Index (DMI)



A trend-following indicator gauging the strength & direction of potential trend reversals

- When there is not a visible trend
- Whichever DI is on top is the current direction of the trend
- A higher ADX shows a stronger trend
- -DI and +DI cross over with an upwards sloping ADX



Coppock Curve

10 WMA of (ROC14 + ROC11)

- Long term momentum
- Above o indicates positive momentum
- Below o indicates negative momentum
- Indicates strength of momentum too





Fibonacci Retracements

A tool for spotting key support and resistance levels using Fibonacci ratios

- Shows the potential most important level
- Highlight potential entry points
- Based on Fibonacci ratios (23,6%;38,2%;61,8%....)



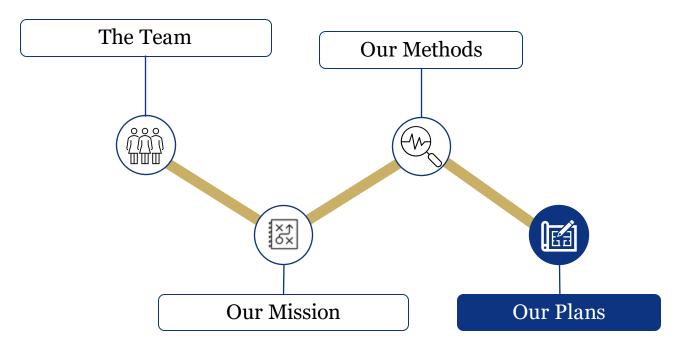
Non-Visual Indicators



Another view at the same problem

- Put/call ratio
- Short interest ratio
- Volume Weighted Average Price (VWAP)
- Advance-Decline ratio





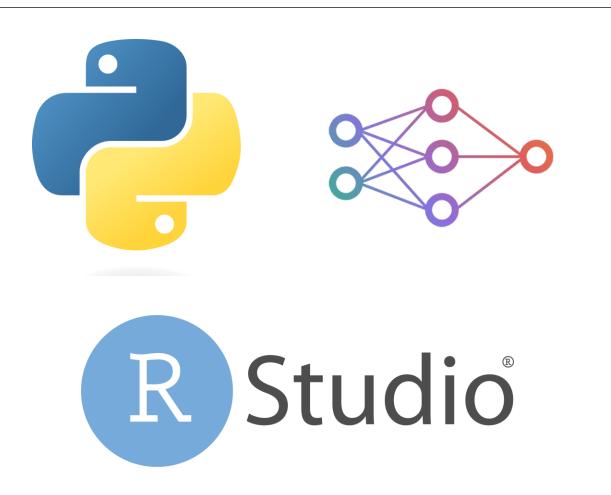


Future Projects

Machine Learning, Modeling & Statistics

We have several exciting plans in the pipeline

- Neural Network for time series prediction
- News headlines for sentiment analysis
- Bayesian Econometrics (Markov chain)
- Random Forest Classifier



Neural Network for Time Series Prediction

META has created the framework for a model

- Deep learning technique non-parametric
- Uses autoregressive integrated moving averages (ARIMA)
- Interesting for analysing trends
- Certain assumptions have to be met explore further

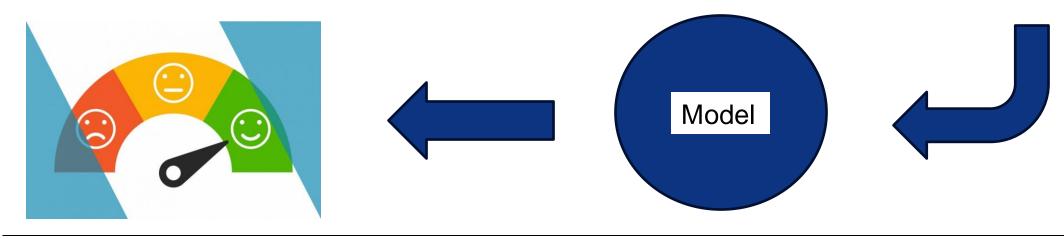


News headlines for sentiment analysis

Sentiment analysis

- News headlines act as data
- Sentiment models can score the headline (Positive or Negative)
- The market sentiment score for a stock can be calculated
- Decide on scoring method
- Include other input sources

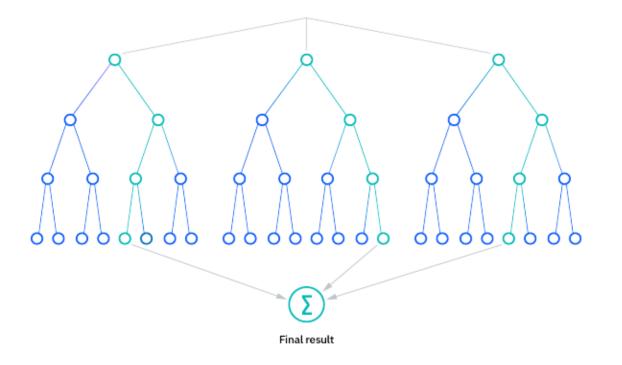
Nvidia is on its way to becoming the first \$2-trillion chipmaker The AI chipmaker's stock soared to a record high after earnings and then continued climbing



Random Forest Classifier

Machine Learning for Stock Return

- Model to identify a high performer stock based on its current fundamentals
- Random Forest Classifier trained on this data
- Produces a Random Forest Score for a stock, indicating the probability that it is a high performer

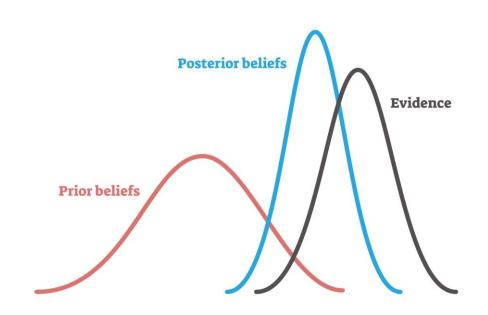


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Bayesian Econometrics

Investigating turning points

- Logistic regression with time series data
- Prior beliefs (expectations)
- Hamiltonian Monte Carlo
- NUTS (No-U-Turn Sampler) algorithm
- Burn-in period equal to 0.25 or 0.5 number of iterations for each chain









Sustainability



The Team



- Yves Baljet
- MSc Sustainable Finance
- Senior Analyst



- Hannes Steinhoff
- BSc International Business
- Junior Analyst

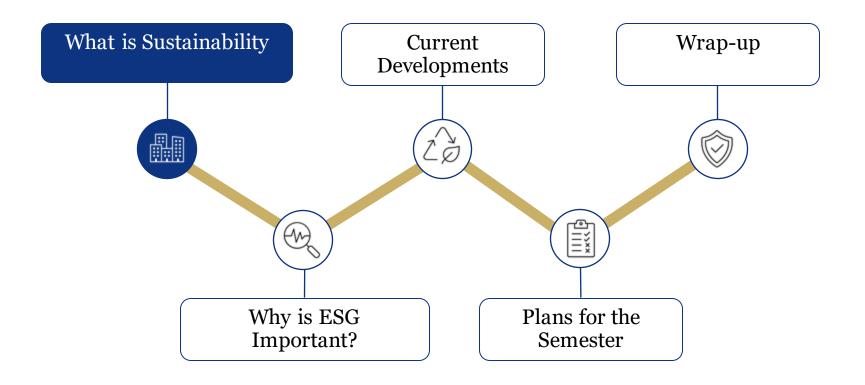


- Vadim Baliakin
- BSc Business Engineering
- Junior Analyst



- Matthias Sciollintano
- MSc Corporate Finance
- Junior Analyst







What is Sustainability

What are the ESG implications for Sigma

Definition and Objectives

- Balancing economic, environmental, and social needs without compromising future generations.
- Preserve natural resources and maintain ecological balance.
- Key to identifying risks and opportunities affecting long-term returns.
- Integrates ethical, environmental, and social considerations
 into investment decisions.

Materiality

- Identifies ESG issues most likely to impact financial performance.
- Helps investors focus on critical sustainability factors.
- Ensures investments align with long-term value creation.

ESG

- Environmental: Climate change, resource depletion, waste management, and pollution.
- **Social**: Labor practices, human rights, community impact, and customer satisfaction.
- Governance: Board diversity, executive pay, ethics, and transparency.
- Help identify potential risks and opportunities.

Positive Vs. Negative Screening

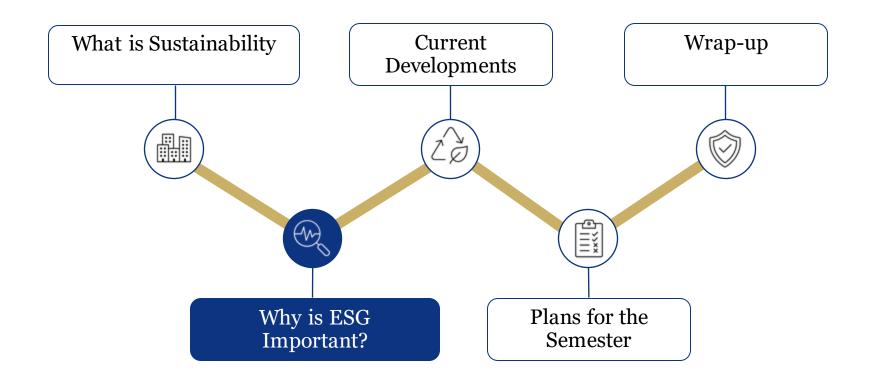
What are ESG implications for Sigma

Screening Objectives

- Filters investments based on ESG criteria.
- Includes firms with strong ESG practices or sectors
- Excludes companies or sectors with negative impacts on society or environment (e.g., fossil fuels, tobacco).
- Shapes investment choices, aligning with ethical and sustainability goals.
- Enhances long-term returns by avoiding companies with high ESG risks.

Implications for Sigma

- Aligns fund with future-oriented, responsible investment strategies.
- Positive Screening: Leads to investments in innovative and sustainable industries, potentially increasing long-term returns.
- Negative Screening: Mitigates risks by avoiding investments in controversial sectors or companies with poor ESG performances.
- Builds the fund's reputation as an ethical and socially responsible investor.





Risks & Opportunities of ESG investing

Risks

ESG risks are social, environmental, and governance factors that have an impact on the financial success and management of a company.



Risks & Opportunities of ESG investing

Opportunities

ESG opportunities are benefits which a company gains in case of caring about its ESG risks.

Financial and Operational Benefits

- Reduced Operational Costs
- Improved Financial Performance
- Access to Capital
- Risk Mitigation
- Regulatory Compliance

Strategic and Stakeholder Benefits

- Enhanced Reputation
- Innovation and Market Opportunities
- Employee Engagement and Retention

Example of ESG influence.

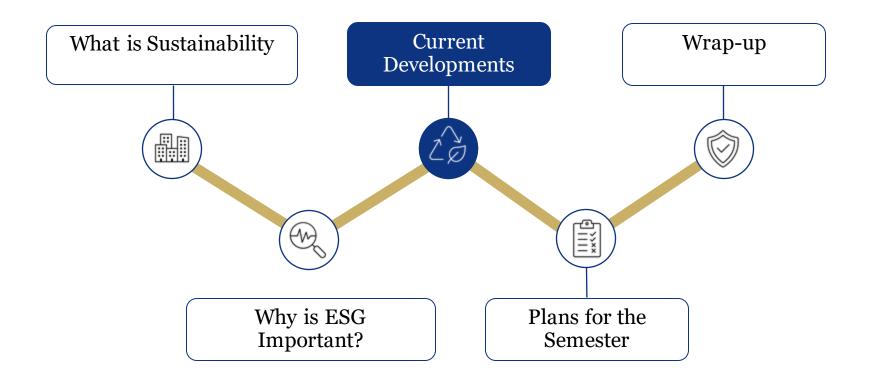
Bayer Monsanto case



Case Description

An ESG risk example from Bayer AG's situation is the extensive litigation surrounding Roundup, alleging its link to cancer. This represents environmental risks due to product safety concerns, social risks impacting public health and affected communities, and governance risks related to legal management and shareholder value protection.



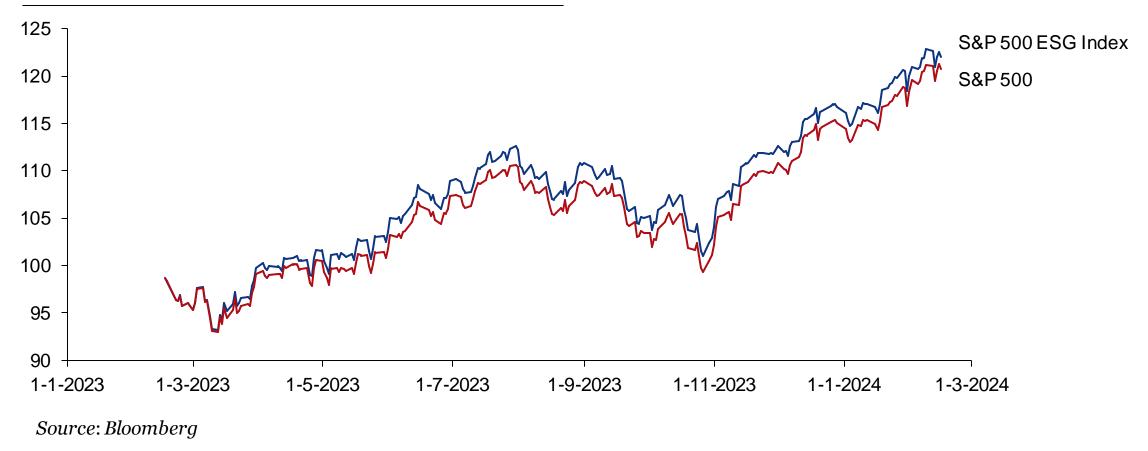




Current Developments in ESG investing

"Green is Good: Wall-Street's new mantra" ~ Financial Times (2021)

S&P ESG Index outperforming S&P500



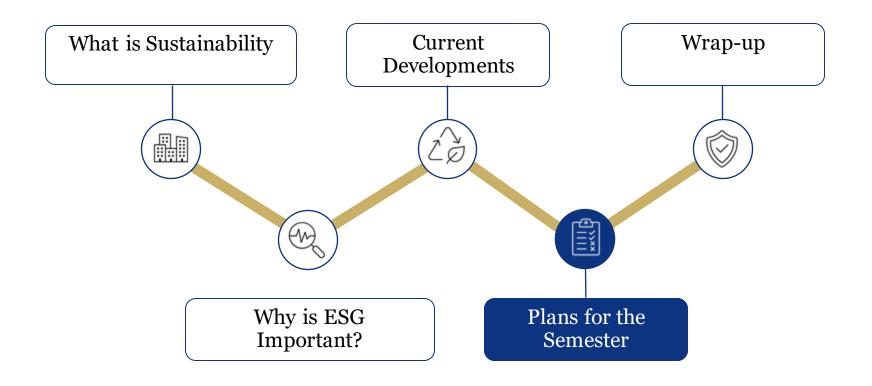
What is happening in 2024

CSRD, Greenwashing, Climate Risks and Supply-Chain Adaptation

Most important topics for this year:

- ESG backlash
- Wait-and-see approach regarding regulation
- Corporate Sustainability Reporting Directive (CSRD) 2024, expanding regulation to more (private) companies
- Increased regulatory action and legislation against greenwashing

- (Physical) Climate Risks increasingly relevant
- Focus on adaptation and resilience planning
- Supply-chain adaptation and integration, measuring Scope
 1/2/3 emissions across the supply chain
- Challenges and opportunities related to A.I.





Plan for the semester

How do we integrate sustainability into Sigma

- Complex ESG Engine
- Manual input
- Combining several agency ESG ratings with own estimates
- Numbers, numbers, numbers....
- No useful qualitative information for analysts and Sigma's investors
- Pitches not interesting to listen to

New Approach

Quantitative

- Create new ESG Engine from scratch
- Code from FactSet: automatic and instant monitoring
- Analyze ESG rating of Sigma's Portfolio
- Benchmark investment pitches

Qualitative

- Tell the story behind the numbers
- Provide useful company-specific information regarding scandals, risk, opportunities etc.
- Making pitches meaningful

Example

Quantitative

Top 5 ESG Risk Exposures



MSCI ESG Rating





Qualitative

Lufthansa: Planetary protection posturing

An advertising campaign for Lufthansa was banned in the United Kingdom for making misleading claims about the German airline's efforts to protect the planet. Britain's Advertising Standards Authority (ASA) said there were no commercially viable technologies in the aviation industry which would substantiate Lufthansa's claim that it was protecting the world's future.

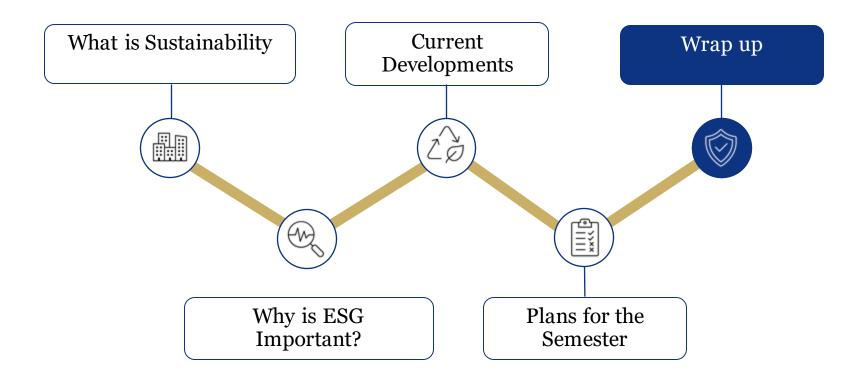
In its defence, Lufthansa – which has pledged to be carbon neutral by 2050 and half its emissions by 2030 – said the tagline was "open to interpretation" and argued that consumers would not see it as an "absolute promise" to protect the planet or that its planes did not cause harm. The ASA said brands in high carbon emitting sectors shouldn't make claims that give consumers a false impression about plans they can't substantiate.



LUPTHANKA GROUP

Source: Eco-Business

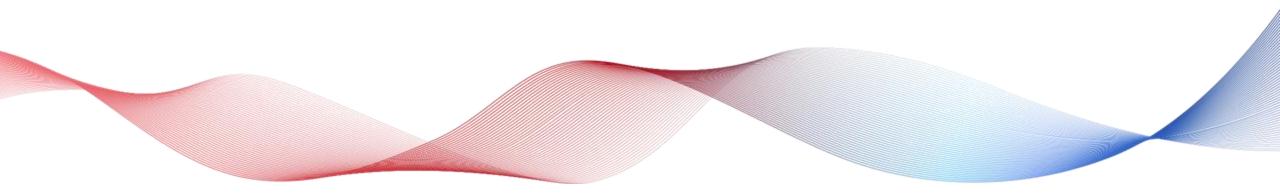
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We need your input!









Mission





Restructuring

New Portfolio Dashboard and ESG Engine

Portfolio Dashboard

- \rightarrow Total clean up
- FactSet and Bing integration
- Metrics sheet
- Budget Plan
- + Report integration

ESG Engine

Mix of hardcoded and manual inputs

- Inefficient
- Unintelligible
- → Revamp engine

Global Fund Approach

Reporting

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Monthly fund and macro report for communication



Voting

Overhaul voting system to increase investment attractiveness within the fund

Current System

- ♦ €500 Minimum Buy-in
- 1 vote per share capped at 3 votes
- No payout Restrictions

Objective

- I. Boost the appeal of investing for students
- II. Establish an efficient voting system to secure higher voter participation

Marginal Approach

- Allows students to invest in increments of €100.
- ◆ Each €100 investment grants one vote, up to a maximum of 15 votes.
- < €300 invested capital
 → 18 month holding period

Tier System

Tier I €500 - 2 Votes No minimum holding period

Tier II €250 - 1 Vote Minimum holding period: 1 year

Tier III €100 - no voting rights Minimum holding period: 2 years

Open Discussion

Do they address the problem? Are they feasible? What are the risks? And what are the alternatives?

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Attendance

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Feedback

