# Explainability Index (EI) \& Risk of Target (RoT) 

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## Picture of the Day



## Quote of the Day



Successful investing is about managing risk, not avoiding it.

Benjamin Graham, father of value investing

## Table of Contents

(1) Introduction
(2) EI Calculation Procedure

## Introduction (1 of 2)

- 100s of proposed performance measures (ranging from very simple to more advanced, measured in different scales, linear/non-linear, etc) are used to खassess securities, evaluate portfolios, create asset allocation profiles, capital adequacy/efficiency, riskmanagement and so on (e.g., return, VaR, Sharpe, Calmar, etc)
- assessments hinge on the relative range of individual performance measures, and usage is based on some form of a grid of select measures with associated weights


## Introduction (2 of 2)

- we propose a Unifying Framework of Performance Measures as an Explainability Index (EI) that captuires the multi-dimensionality and nuances measured by the individual measures, where it balances the different input categories of performance measures according to default or specified preferences and gives acomposite bounded score between 0 and 1.
- we also propose a relative measure as the Risk of Target (RoT) that leverages the EI for comparing the performance of assets/portfolios/etc with their targets and assesses the drivers of divergence.


## Table of Contents

(2) EI Calculation Procedure
(3) Practical Appliceoions

## Scaling

- first realization is that performance measures do not have the same scale
$\triangleright$ making their direct comparison non-practical
- $1^{\text {st }}$ step: scaling of performance measures
$\triangleright$ uše a sigmoid function to do so (scale values between 0 \& 1)


## Scaling - Sigmoid

- a separate issue rises with the sigmoid function; given the nature of the value of the features, some performance measures get mapped into bucketed regions
- e.g. Batting Average has arange of values between $0 \%$ and $100 \%$, resulting in a value between 0.5 and 1 after applying a sigmoid function ${ }^{1}$
- on the other hand, Max Drawdown has a range between $-100 \%$ and $0 \%$, resulting on a value between 0 and 0.5
- therefore, cannot compare scaled values of these performance measures

[^0]
## Two-step transformation (1 of 2)

- for the sigmoid transformation to properly scale the measures, we apply a linear transformation beforehand
- linear mapping that aligns $\alpha \times 100 \%$ \& ( 1 er $\alpha) \times 100 \%$ of sigmoid with corresponding historicaldistribution respectively
- required values from the sigmoid function are obtained by:

$$
\frac{e^{c} 1}{1+e^{-x_{\alpha}}}=\alpha
$$

solving for $x_{\alpha}^{\text {e }}$ to get

$$
x_{\alpha}=-\ln \left(\frac{1-\alpha}{\alpha}\right)
$$

and

$$
x_{1-\alpha}=\ln \left(\frac{1-\alpha}{\alpha}\right)
$$

## Two-step transformation (2 of 2)

- the linear transformation is given by

$$
\hat{m}=\beta\left(m-m_{\alpha}\right)-x_{\alpha}
$$

where

$$
\beta=\frac{x_{1-\alpha}-x_{\alpha}}{m_{1}-\alpha-m_{\alpha}}
$$

and $m_{\alpha}$ and $m_{1-\alpha}$ arefrom historical distribution

- for this step cvalues could come from the asset, the index, or a pool of multiple assets/indexes
thus we obtain

$$
\tilde{m}=\frac{1}{1+e^{-\hat{m}}}
$$

## Transformation Example




Linear Transformation

$$
\beta=\frac{x_{1-\alpha}-x_{\alpha}}{m_{1-\alpha}-m_{\alpha}}
$$

Where $\beta$ is slope, mapping:

$$
\hat{m}-x_{\alpha}=\beta\left(m-m_{\alpha}\right)
$$

With $m$ as original, and $\hat{m}$ as transformed via linear mapping


## Proper Orientation of Performance Measures

- when building an index, we need to take into account the importance of the performance orientation
- for performance measures like Volatility the lower the better and for Return the higher the better
- by construction, EI assumes 0 is better than 1 , so direction needs to be adjusted accordingly
- in case of the higher the better, define $\bar{m}=1-\tilde{m}$, otherwise $\overline{\mathrm{m}}=\tilde{m}$
- after this adjustment, all performance measures have the same scale and orientation


## Transformation Example - Return



Linear \& Sigmoid - Return for Large Cap Mutual Funds on 2021-12-3 Linear Transformation


## Transformation Example - Volatility




## Transformation Example - Max Drawdown

Sigmoid - Max Drawdown for Large Cap Mutual Funds on 2021-12-31



Linear \& Sigmoid - Max Drawdown for Large Cap Mutual Funds on 2021-12-31


## Transformation Example - Sharpe Ratio

Sigmoid - Sharpe Ratio for Large Cap Mutual Funds on 2021-12-31
Original Measure

 Original Measure




## Categories

- in order to enhance explainability of EI, we define categories based on user preferences
- four default categories are: return, volatility, drawdown, and alternatives
- within each category, all transformed performance measures are equally weighted which yields a number for each category
- when it comes to combining categories, users can use their own weights based on their preferences (default is equally weighted)


## Explainability Index (EI)

- to unify all categories into a single value, i.e. EI, users can employ one of the below methods:
- arithmetic: $\sum_{k=1}^{K} w_{k} \bar{m}_{k}$
- geometric: $\prod_{k=1}^{K} \bar{m}_{k}^{w_{k}}$
- distributional: 1 -
- $d_{H}$ is Hellinger distance
- $H^{2}\left(\mu_{1}, \mu_{2}, \sigma_{1}, \sigma_{2}\right)=1-\sqrt{\frac{2 \sigma_{1} \sigma_{2}}{\sigma_{1}^{2}+\sigma_{2}^{2}}} e^{-\frac{1}{4} \frac{\left(\mu_{1}-\mu_{2}\right)^{2}}{\sigma_{1}^{2}+\sigma_{2}^{2}}}$
- $d_{H}=\frac{1}{N} \sum_{i=1}^{N} H^{2}\left(R^{(i)}, R\right)$


## Risk of Target (RoT)

- comparing EI of an asset with EI of its benchmark would allow portfolio managers to assess their performance deviation from the target benchmark i.e. RoT $T_{C}$
- we can calculate RoT as a difference

$$
\mathrm{RoT}=\mathrm{EI}^{\text {asset }}-\mathrm{EI}^{\text {target }}
$$

or as a percentage difference

$$
\text { RoT }=\frac{\mathrm{EI}^{\text {asset }}-\mathrm{EI}^{\text {target }}}{\mathrm{EI}^{\text {target }}}
$$

## User-defined Parameters

- here are user-defined parameters/preferences
- type of performance measures
- level of $\alpha$ for sigmoid transformation
- asset or group of assets used to determine $z_{\alpha}$ and $z_{1-\alpha}$
- categories
- weights
© 'EI methodology
- variation across time ${ }^{2}$


## Table of Contents

(2) EI Calculation Procedure
(3) Practical Applications

## Example 1 - Efficient Frontier

- efficient frontier compares assets in 2D, return \& volatility
- this could create undesirable risks




## RoT Efficient Frontier

- RoT adds color for better visualization purposes
- it improves understanding in higher dimensional hidden risk


$\square$


## RoT Fund Analysis

this is a deeper dive into one of the funds (can be for any asset)


## RoT Multi Fund Analysis: Index Tracking ETFs (1 of 2)

- it makes RoT an ideal tool to compare howaćcurately an ETF can replicate the behavior of an index
- index tracking ETFs should have a nearly identical holding composition to their respective index but affected by fees, taxes, and transaction cost
- given that there are multiple index tracking ETFs, it is crucial to ${ }^{\text {assess }}$ which one follows closely their respective index


## RoT Multi Fund Analysis: Index Tracking ETFs (2 of 2)

Benchmark [S\&P 500] vs ETFs

$k<\measuredangle \square \triangle>\square++$

## Example 2 - RoT of Portfolios (1 of 2)

- RoT is ideal to assess portfolios against target objectives
- it is possible to combine RoT values of individual assets given that it is highly non-linear
- a portfolio RoT requires the calculation of all measures from portfolio monthly returns
- when presented with multiple portfolios that are in proximity to each other in the efficient frontier, RoT can portray their entire risk profile


## Example 2 - RoT of Portfolios (2 of 2)

Benchmark [60\% S\&P 500 40\% US Agg] vs Portfolios ©ak ai

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## Table of Contents



## Conclusion - power of EI \& RoT

- unify processes, assessments, and explanations
- add color and capture nuances (linearforon-linear) in a simple and explainable manner
- compare assets/portfolios in a uniform manner at a point in time (relative or trend)
- construct multi-objective asset allocation profiles/portfolios
- extend to incorporate any/all performance measures, alternative measures (e.g., Expense Ratios), holdings, etc


[^0]:    ${ }^{1}$ for volatility would result in a value greater than 0.5

