Explainability Index (EI) & Risk of Target  $(Rot)^{2023}$ Ence on April 15

# Ali Hirsa

Professor & Director of Financial Engineering Industrial Engineering & Operations Research (IEOR) Data Science Institute (DSI) Director of Center for AT in Business Analytics & FinTech Presentation at Princeton Columbia University Chief Scientific Officer, ASK2.ai

Managing Partner, Sauma Capital, LLC

Presentation at Princeton FinTech & Quant Conference, April 15<sup>th</sup>. 2023 (日本本語を本書を本書を入事)の(で)

# Picture of the Day



Practical Applications

#### Quote of the Day



Successful investing is about managing risk, not avoiding it.

Benjamin Graham, father of value investing

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

Practical Applications

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 - のへで

#### Table of Contents

	April 15th, 201
1 Introduction	* Conference ou
2 EI Calculation Procedure	tuant
3 Practical Applice or Brind	
A Cation Presentation	

# 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 assess securities, evaluate portfolios, create asset allocation profiles, capital adequacy/efficiency, risk management 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)

- ril 15th , 202 • we propose a Unifying Framework of Performance Measures as an Explainability Index (EI) that captures 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 accomposite bounded score between 0 ceton Fin 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.

Practical Applications

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 - のへで

#### Table of Contents

	April 15th, 20
1 Introduction	conference on '
2 EI Calculation Procedure Quant	
3 Practical Applice OBns	
Cation at Presentation	

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

# Scaling

- first realization is that performance measures do not have the same scale
   making their direct comparison
  - FinTech

Prince
 use 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 a range of values between 0% and 100%, resulting in a value between 0.5 and 1 after applying a sigmoid function<sup>1</sup> Fin<sup>Tech</sup>
- 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

<sup>&</sup>lt;sup>1</sup> for volatility would result in a value greater than  $0.5 \rightarrow (2) \rightarrow (2$ 

# Two-step transformation (1 of 2)

- for the sigmoid transformation to properly scale the measures, on April we apply a linear transformation beforehand
- linear mapping that aligns  $\alpha imes 100\%$  &  $(1 \not\in \alpha) imes 100\%$  of sigmoid with corresponding historical distribution respectively
- required values from the signoid function are obtained by:

solving for 
$$x_{\alpha}^{\text{eton}}$$
 to get  
 $x_{\alpha} = -\ln\left(\frac{1-\alpha}{\alpha}\right)^{2}$ 

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

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

Introduction

EI Calculation Procedure

Practical Applications

Conclusion

#### Two-step transformation (2 of 2)

the linear transformation is given by

on (2 of 2)  
ation is given by  

$$\hat{m} = \beta(m - m_{\alpha}) - x_{\alpha}$$
 on April 15<sup>th</sup>, 20  
 $\beta = x_{1-\alpha} - x_{\alpha}$ 

where

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

and  $m_{\alpha}$  and  $m_{1-\alpha}$  are from historical distribution

• for this step, values could come from the asset, the index, or a pool of multiple assets/indexes

Presso tation at manufi

$$ilde{m} = rac{1}{1+e^{-\hat{m}}}$$

# Transformation Example



# 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  $\bar{m} \stackrel{m}{\cong} \tilde{m}$ 
  - after this adjustment, all performance measures have the same scale and orientation

Practical Applications

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

# Transformation Example - Return



Practical Applications

# Transformation Example - Volatility





▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 のへで

Practical Applications

Conclusion

#### Transformation Example - Max Drawdown





Practical Applications

Conclusion

#### Transformation Example - Sharpe Ratio





◆□▶ ◆□▶ ◆三▶ ◆三▶ - 三 - のへで

# 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)

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

#### Explainability Index (EI)

- to unify all categories into a single value, i.e. EI, users can employ one of the below methods: • geometric:  $\prod_{k=1}^{K} \bar{m}_{k}^{w_{k}}$ • distributional:  $1 - (1 - EI) \times (1 - d_{H})$ •  $d_{H}$  is Hellinger distance •  $H^{2}(...$

• distributional: 
$$h^{-1}(1 - EI) \times (1 - d_{H})$$
  
•  $d_{H}$  is Hellinger distance  
Presentation  
•  $H^{2}(\mu_{1}, \mu_{2}, \sigma_{1}, \sigma_{2}) = 1 - \sqrt{\frac{2\sigma_{1}\sigma_{2}}{\sigma_{1}^{2} + \sigma_{2}^{2}}}e^{-\frac{1}{4}\frac{(\mu_{1} - \mu_{2})^{2}}{\sigma_{1}^{2} + \sigma_{2}^{2}}}$   
•  $d_{H} = \frac{1}{N}\sum_{i=1}^{N}H^{2}(R^{(i)}, R)$ 

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

# 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<sub>C onf</sub>er
- we can calculate RoT as a difference
  - FROT = EI<sup>asset</sup> EI<sup>target</sup>

or as a percentage difference Presentation at Presentation at Presentation Presentation

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

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

#### User-defined Parameters

- -es on April 15th, 202 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  $z_{1-\alpha}$
  - sories weights ceton Finte
- PresentationEI methodology • variation across time<sup>2</sup>

Practical Applications

・ロト ・ 国 ト ・ ヨ ト ・ ヨ ト

æ

#### Table of Contents



Practical Applications 0000000

(a)

э

# Example 1 – Efficient Frontier

- ril 15th, 204 efficient frontier compares assets in 2D, return & volatility
   this could create undesirable risks



# **RoT Efficient Frontier**

- RoT adds color for better visualization purposes
- wil 15th , 204 • it improves understanding in higher dimensional chidden risk nce





us\_agg





Practical Applications

#### **RoT Fund Analysis**

this is a deeper dive into one of the funds (can be for any asset) . any as April April Presentation at Princeton FinTech & Quant Conference on April

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

- it makes RoT an ideal tool to compare how accurately an ETF can replicate the behavior of an index of the second se
- 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 assess which one follows closely their respective index

Practical Applications

Conclusion

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

# Presentation at Princeton FinTech & Quant Conference on April 15th, 201 RoT Multi Fund Analysis: Index Tracking ETFs (2 of 2)

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

# 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

Practical Applications

# Presentation at Princeton FinTech & Quant Conference on April 15th, 204 Example 2 – RoT of Portfolios (2 of 2)

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

Practical Applications

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 - のへで

#### Table of Contents

	April 15th, 20
1 Introduction	conference on ra
2 EI Calculation Procedure	Quant CC
3 Practical Applice or Brinder	
Caniclusion Presentanclusion	

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

#### Conclusion - power of EI & RoT

- unify processes, assessments, and explanations on April 15<sup>th</sup>, 20
- add color and capture nuances (linear/non-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