

See discussions, stats, and author profiles for this publication at: <https://www.researchgate.net/publication/389953574>

Online probabilistic estimation of carbon beta and carbon Shapley values for financial and climate risk

Poster · March 2025

CITATIONS

0

READS

75

1 author:



Thierry Moudiki

39 PUBLICATIONS 11 CITATIONS

SEE PROFILE

Beta, Carbon beta, Carbon Shapley value

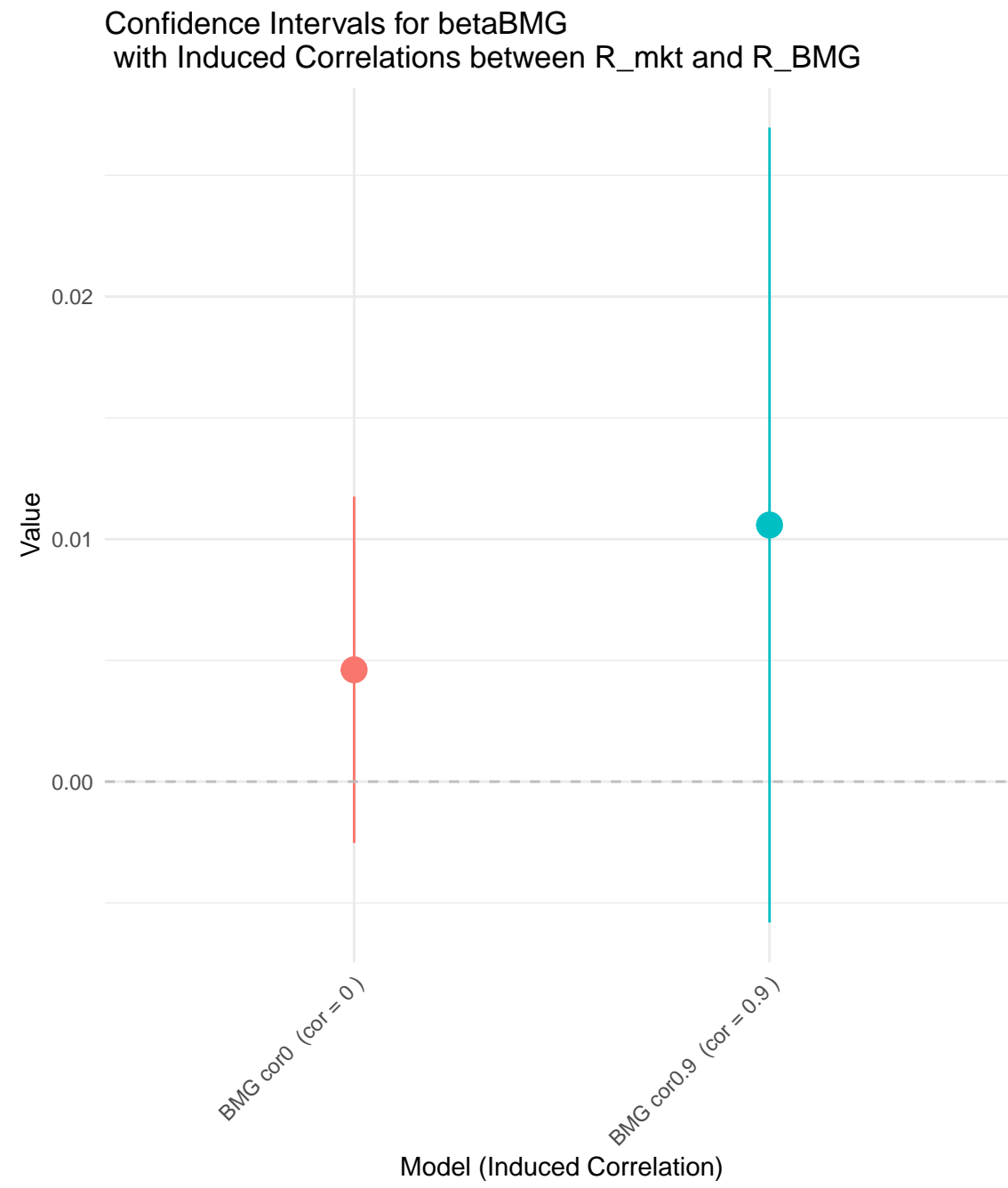
- $R_i(t)$ = Return of individual stock i at time $t > 0$
- $R_{mkt}(t)$ = Return of the global stock market (e.g CAC40, DAX, ...) at time t
- α_i = Expected return of individual stock i when market return == 0
- $\epsilon_i(t)$ is the model's error term for stock i at time t
- $\beta_{mkt,i}$ = **beta for stock i**

$$\text{Beta} : R_i(t) = \alpha_i + \beta_{mkt,i} R_{mkt}(t) + \epsilon_i(t)$$

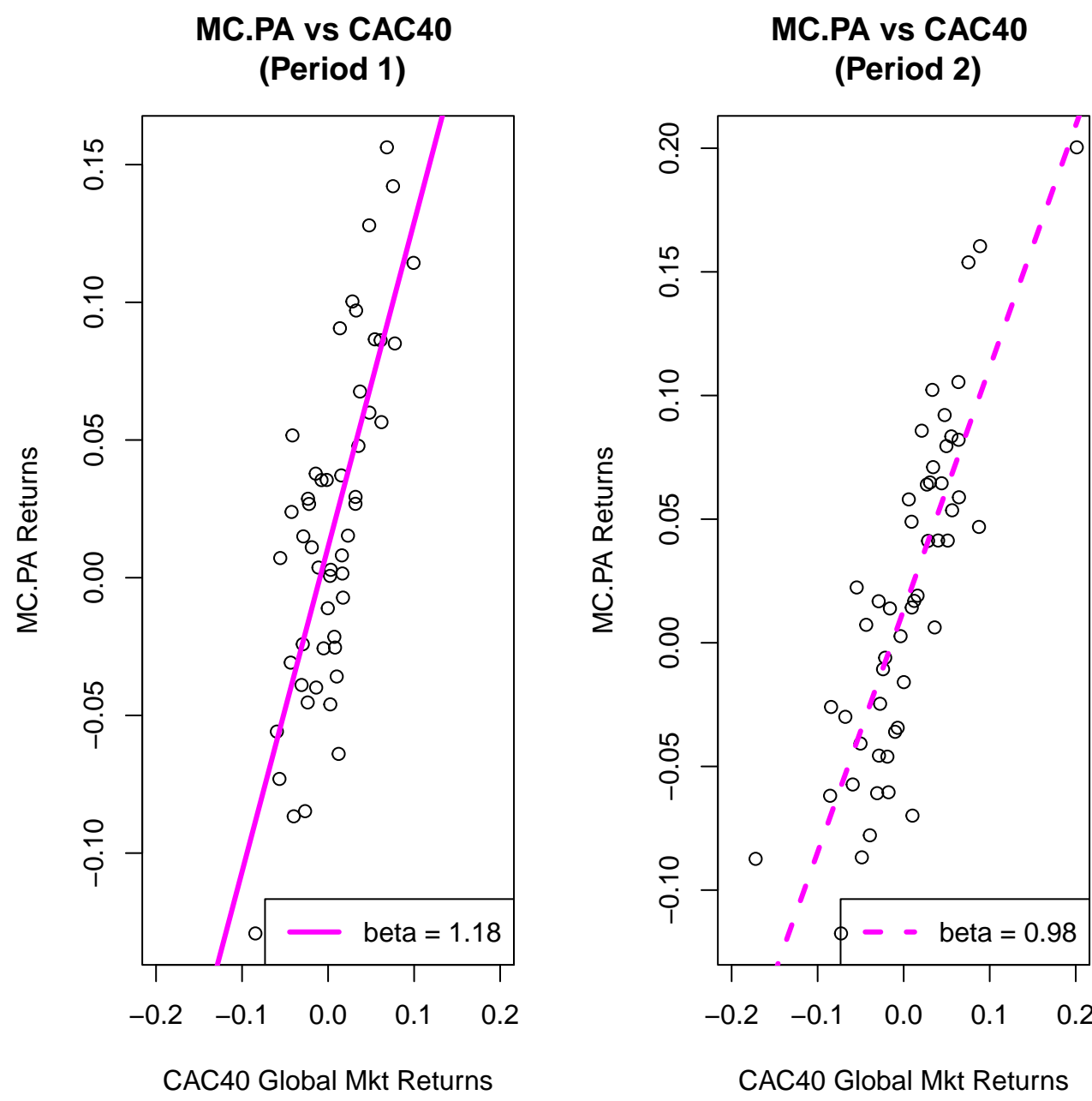
- $R_{BMG}(t)$ = return of **Brown Minus Green (BMG)** portfolio (long Brown, short Green)
- Brown = Less Climate Friendly stocks
- Green = More Climate Friendly stocks
- $\beta_{BMG,i}$ = **Carbon beta for stock i**
- high $\beta_{BMG,i}$ → greener economy will make stock value decrease

- Remarks
- Assumption: **linear** relationships between returns
 - "Extreme" case: **Multicollinearity**: potential high variance for $\beta_{mkt,i}$ and $\beta_{BMG,i}$ estimation

$$\text{Carbon Beta} : R_i(t) = \alpha_i + \beta_{mkt,i} R_{mkt}(t) + \beta_{BMG,i} R_{BMG}(t) + \epsilon_i(t)$$



- $\beta_{mkt,i}$ and $\beta_{BMG,i}$: **not constant** and random

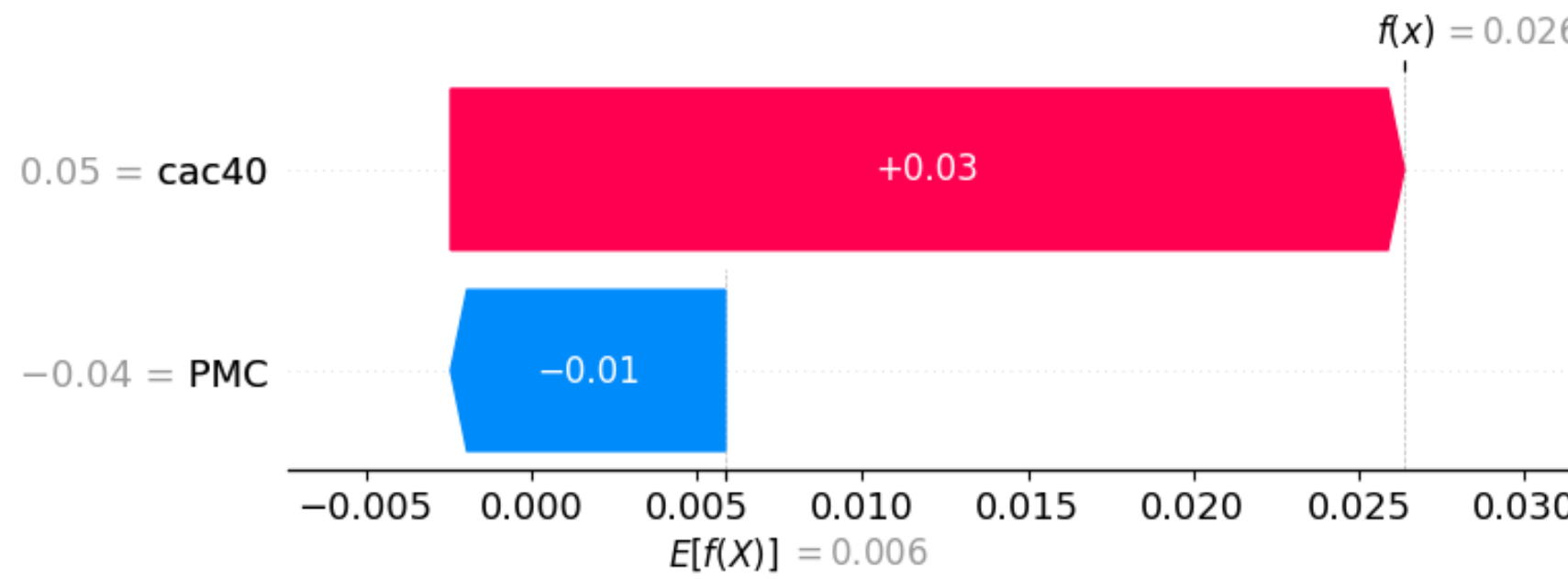


- Paper: Uses a prospective Machine Learning-based (ML) estimation with numerical derivatives = β sensitivities.
- Paper: Compares **Random Forest** model, Rolling **Linear** model, **Random Vector Functional Link (RVFL)** networks model.
- Paper: **Random Vector Functional Link (RVFL)** networks model = (non)linear and regularized + **Online learning (update model as observations arrive)**
- Paper: **Carbon β = numerical derivative of ML model prediction $\hat{R}_i(t)$ relative to $R_{BMG}(t)$ changes**
- Paper: In addition to **Carbon β** , uses **Carbon Shapley values** in ML
- Paper: Shapley values for BMG factor = how much $\hat{R}_i(t)$ changes when $R_{BMG}(t)$ joins different coalitions (subsets of other factors)
- Paper: **Cool thing about Shapley values: additivity no matter the ML estimation model**, i.e

$$\hat{R}_i(t) = \hat{R}_i + \text{Shapleyvalue}_{mkt} + \text{Shapleyvalue}_{BMG}$$

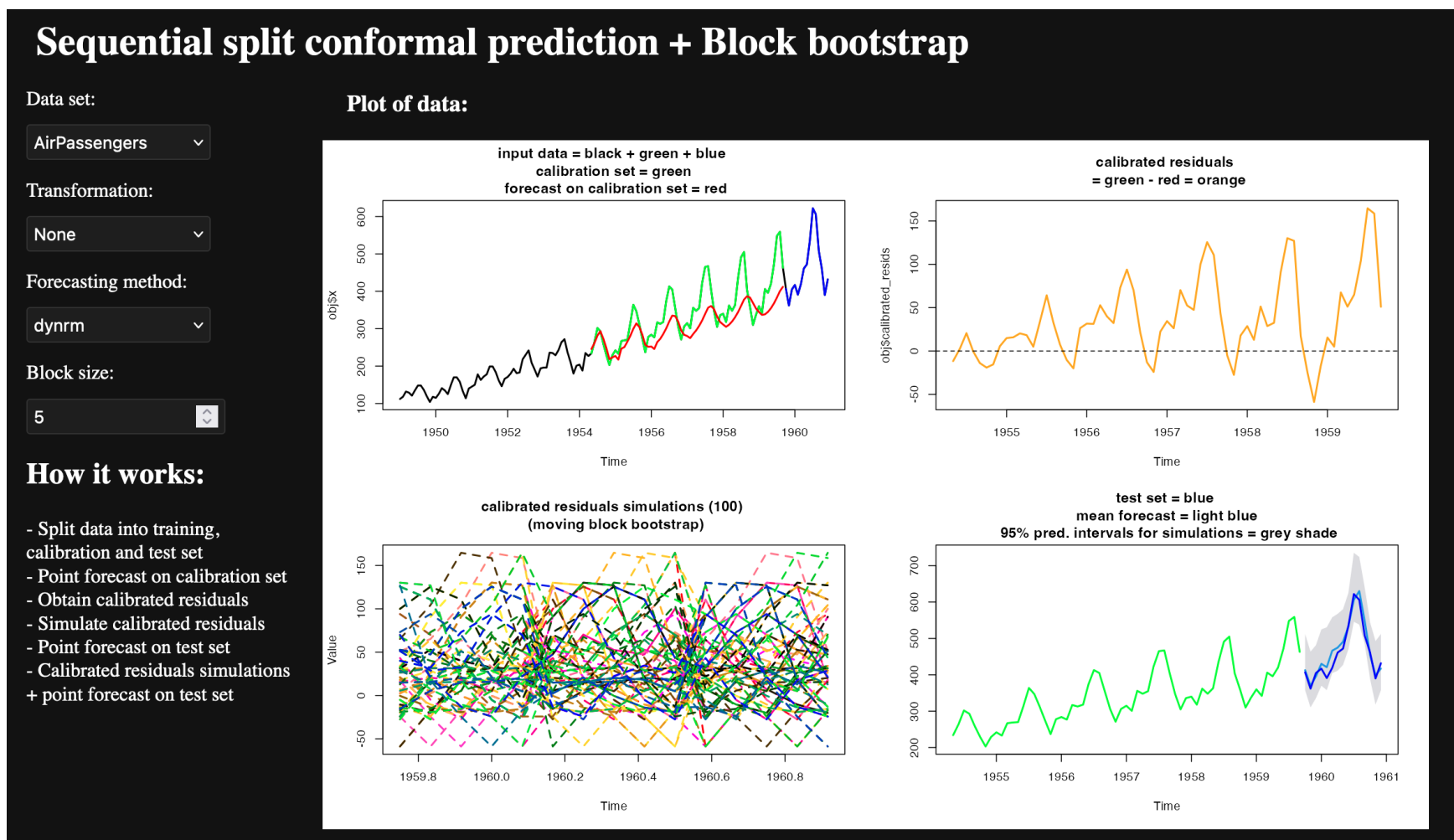
In layman terms:

Predicted return stock i = Average historical return for stock i on a given period + Market contribution to return for stock i + BMG portfolio contribution to return for stock i



In this specific example: **transitioning to a lower carbon economy** would have a positive effect on average stock return.

- Uncertainty around **Carbon β** quantified via **sequential split conformal prediction**.



(1) Results

(2)

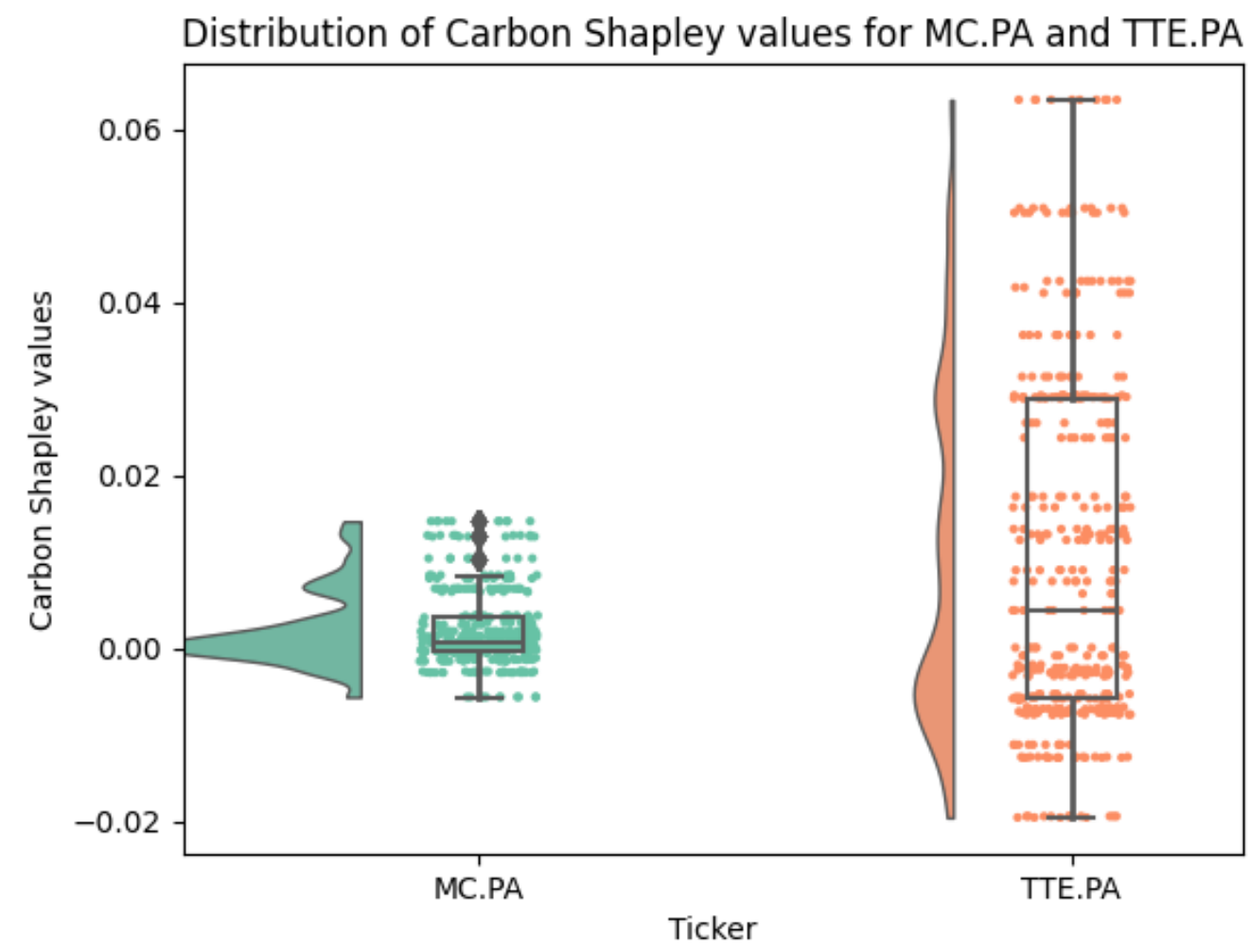
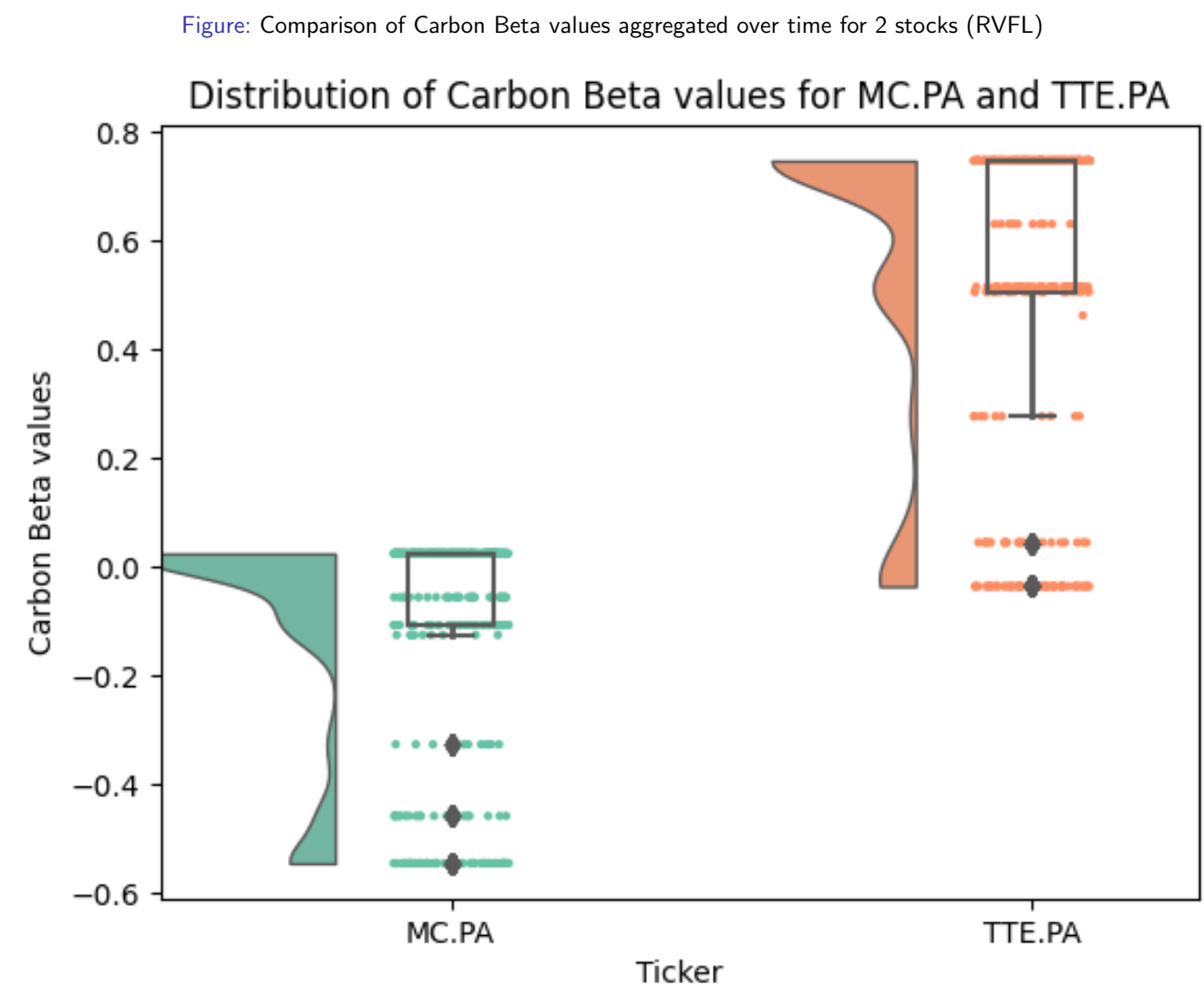
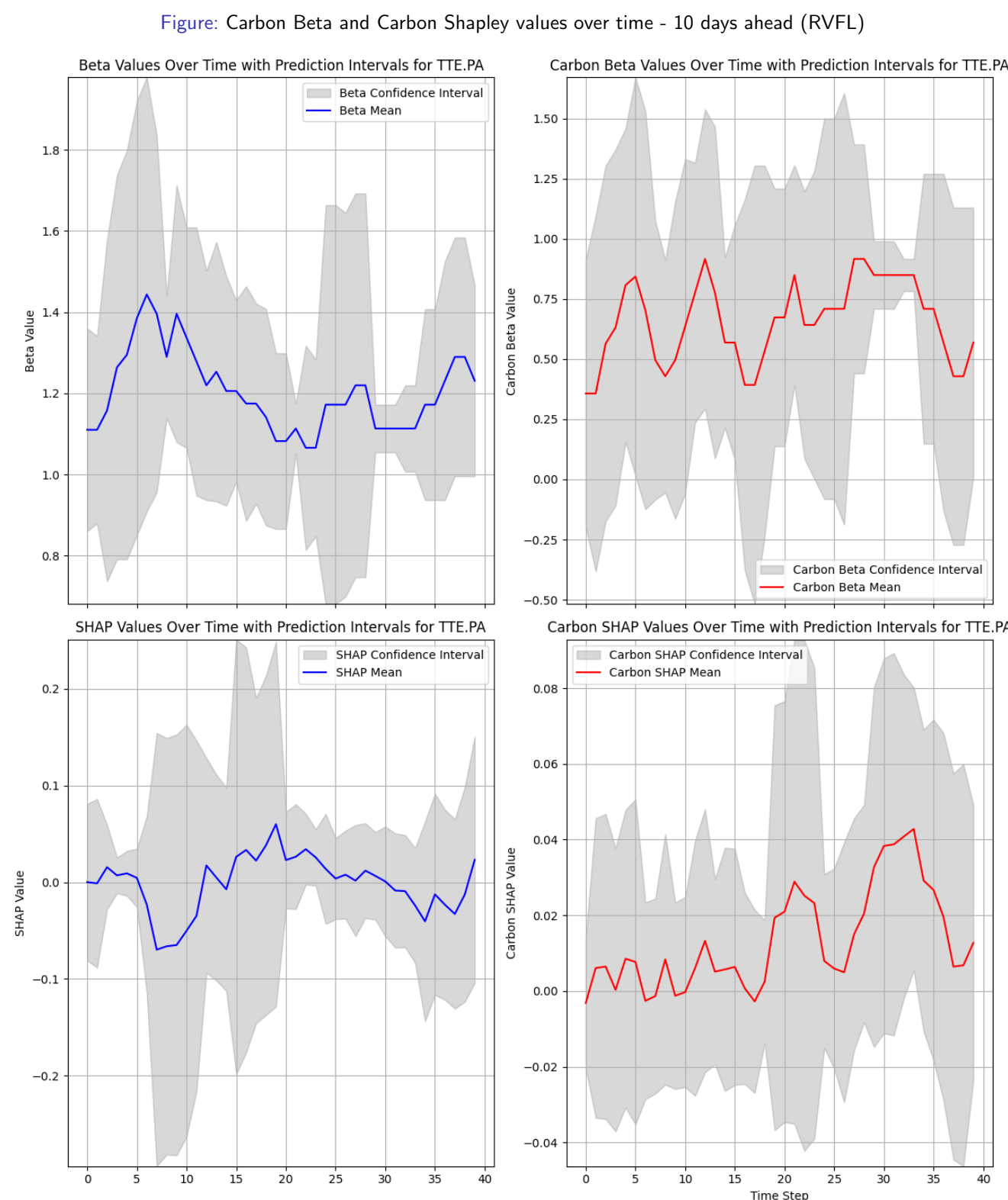
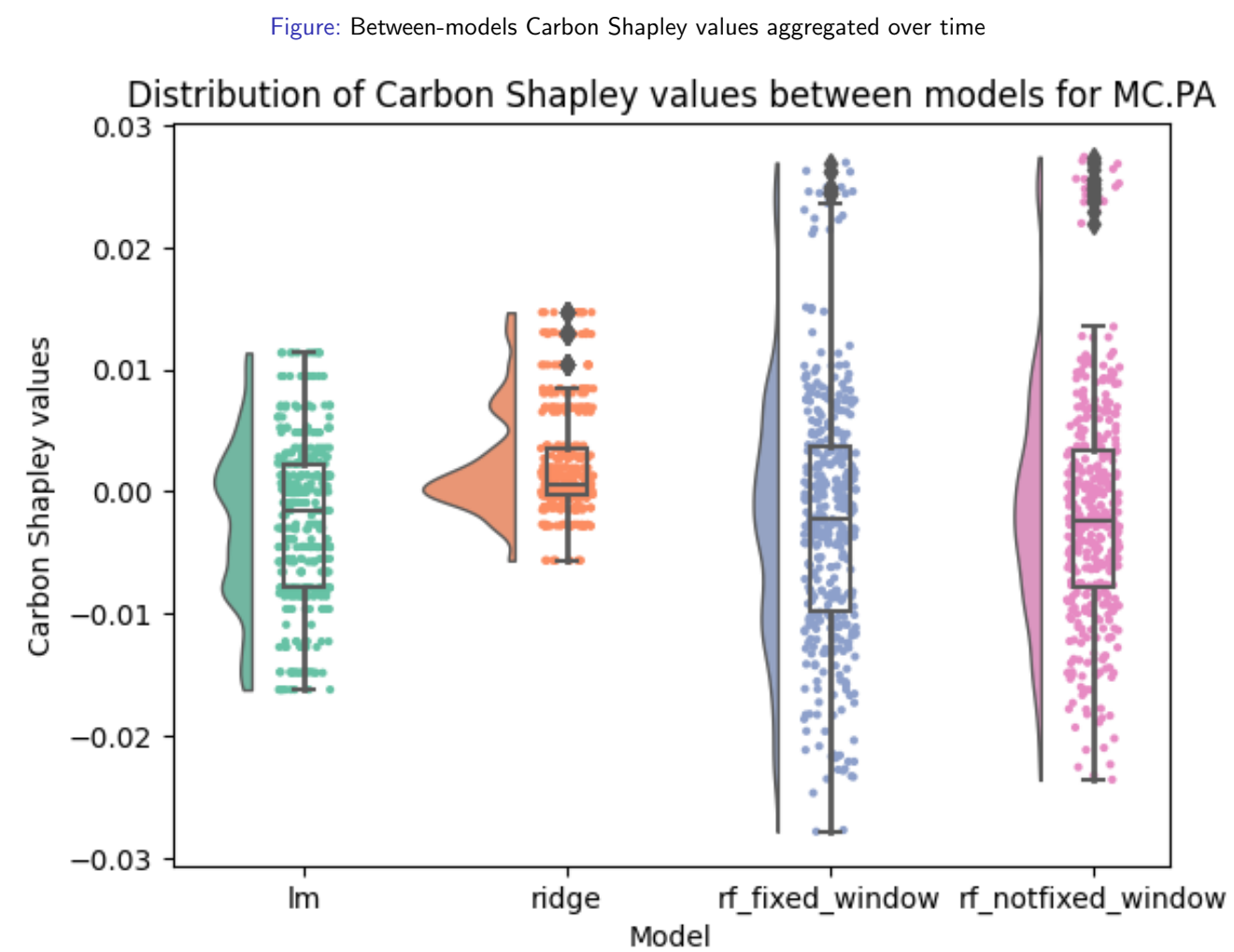


Figure: Comparison of Carbon Shapley values aggregated over time for 2 stocks (RVFL)

Conclusion

- Novel estimation approach for Carbon Beta and introducing Carbon Shapley values.
- More extensive experiments on more stocks would be useful.
- Would be great to have updated versions of BMG portfolio.

References

- Paper described here: https://www.researchgate.net/publication/387677137_Online_probabilistic_estimation_of_carbon_beta_and_carbon_shapley_values_for_financial_and_climate_risk
- Grgen et al. (2020) - Carbon Risk.
- Roncalli et al. (2020) - Time-varying Carbon Beta estimation.
- Lundberg & Lee (2017) - Shapley values in machine learning.

