

Forecast Adjustment Under Shocks: A Unification

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Abstract

This work systematizes and unifies the rich landscape of model adjustment and model correction methods, with a special focus on forecast adjustment under the presence of shocks.

1 Plan

1.1 What the paper should do

- Introduce an existing diffuse set of approaches to adjusting forecasts
- Explain when/how similarity can help us forecast.
- distinguish the method from various tools that inspired it
- show a few special cases, both examples and formal results
- discuss limitations of the method
- propose extensions

1.2 What the paper should NOT do

- Wade too deeply into distance-based weighting details
- wade too deeply into any of the special cases
- Tackle non-scalar random quantities (density forecasts, etc)
- Forecast combination

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2 Introduction

For various modeling and prediction tasks in time series and panel data, the salient challenge is not predicting when an event will occur but what its key properties will be. In the familiar case of scalar time series, that can include its post-event direction, moments, sign, magnitude, duration, and correlation structure, all over an arbitrary horizon or perhaps multiple horizons. This is not to say that predicting the arrival of an event is easy. In some cases, it may be difficult or impossible, and therefore preparing for an anticipated shock is the best one can hope for.

This work focuses on model adjustment amid structural shocks that undermine the reliability of the model at hand. Forecasting under anticipated shocks raises unavoidable questions: should the forecast model be abandoned in favor of a discretionary or ad-hoc or one-off adjustment? Does the discretion of a forecaster rule out a quantitative method for making the adjustment? What is the ultimate purpose of the adjustment, and how it is to be used? For how long is the adjustment necessary or reliable?

Herein we systematize and unify a range of conceptual approaches and tools that have developed across the broad ecosystem of the econometric and forecasting literatures. Additionally, we delineate a specific type of forecasting task called post-shock forecasting, which we broadly define as forecasting under a known shock.

Forecast model adjustment, known most widely perhaps by the term intercept-correction, has received the most attention in several articles and book chapters [Hendry and Clements, 1994, Clements and Hendry, 1996, 1998]. Of special importance is the distinction between discretionary and automated intercepts corrections. Hendry and Clements [1994] define scalar intercept corrections to be automated when they follow the simple rule of adding an estimation residual e_t to subsequent (possibly but not necessarily all) forecasts $\hat{f}_{t+1}, \hat{f}_{t+2}, \dots$. This procedure can colloquially be referred to as setting the model back on track [Hendry and Clements, 1994].

what we are talking about here is not forecast combination, but there may be, nevertheless, a role for forecast combination: combining the forecasts generated by small differences in covariate and/or donor choice, as is done in Lundquist and Eck [2024].

Bias-variance tradeoff and MSE decomposition

The forecast horizon — does it matter? If so, how so? Lin and Eck [2021] has a one-period horizon. Clements and Hendry [1998] p. 203 discuss how long to keep the forecast adjustment in place. For a corrected “slope parameter”, the effect of h is not so clear.

2.1 Literature Review

2.1.1 Motivation for Intercept Correction

Mismeasured data is discussed in [Hendry and Clements, 1994, p. 166] as a motivator for intercept correction. Could similarity-based correction help?

Quinton-Guerrera and Zhong - concerned with correcting β using similarity

Evaluating a Model by Forecast Performance Clements and Hendry [2005]

1. unconditional versus conditional, models;
2. internal versus external standards;
3. checking constancy versus adventitious significance;

4. ex ante versus ex post evaluation (skip this one?);
5. 1-step versus multi-horizon forecasts – this is a relevant question to ask in the context of post-shocking forecasting: should we correct the earliest forecast and then allow the shock to propagate, or should we just correct each term in the horizon, $h=1, 2, \dots, H$?
6. in-sample fixed coefficients versus continuous updating

2.2 The Meaning and Role of Similarity

Hume?

Quantitative ways of determining similarity

Distance function

Asymmetric distance functions are useful when we want to explore differences between donors and weight different contributors to that function differently.

What about qualitative ways of determining similarity – [Lundquist and Eck \[2024\]](#) donors are identified qualitatively

3 Setting

4 Model Adjustment Using Similarity-Based Parameter Correction: A Global Overview

In this section, we introduce and discuss a particular approach to model adjustment that is motivated by the circumstances laid out in Section 3.

1. a random object to forecast that depends on a linear specification or a specification that can be linearized
2. a parametric model family shared by donors
3. a correction term for the model family shared by donors
4. a parametric specification for the correction term
5. a reliable estimation procedure for the shared model
 - (a) This should be straightforward
6. a reliable estimation procedure for the correction term
 - (a) This might not be straightforward. Some models like GARCH, for example, might deliver very noisy estimates for indicator variables that occur just once.
7. a correction function that aggregates (i.e. maps) donor correction terms based on some notion of similarity

Emphasize that there is a non-parametric version of the above: for example, one can use LSTM to predict each of the donor shocks, and then the residuals (or transformations of those residuals, e.g. to percentages) of those n models can be weighted to arrive at a correction term.

5 Model-Specific Considerations

5.1 ARIMA

5.2 GARCH

5.3 HAR

subsectionVAR

5.4 LSTM/GRU

- a
- b
- c
- d

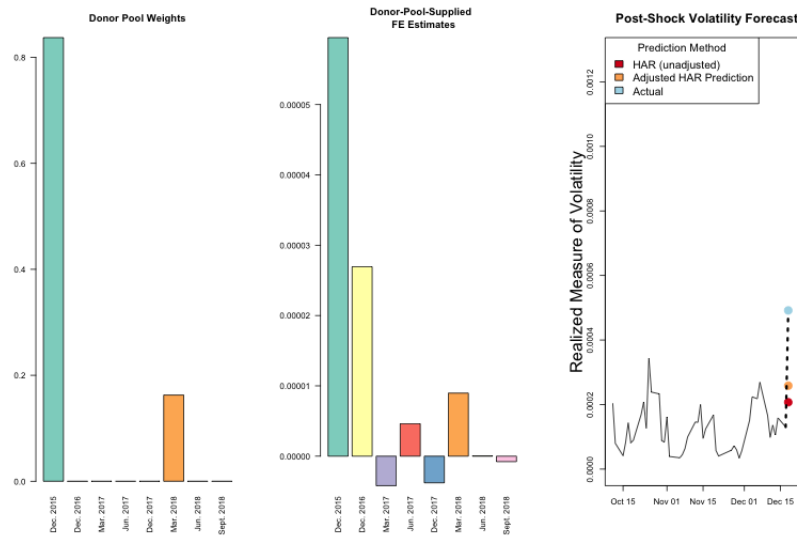


Figure 1: Volatility series of six i.i.d. GARCH processes, each of which experiences a volatility shock, indicated with a red vertical line, at a uniformly distributed point in the set $\{756, \dots, 2520\}$ of trading days, corresponding to between 3 and 10 years of daily trading data.

6 Real Data Examples

7 Discussion

How important is a shared DGP?

Are there any situations where we would want to intercept-correct using the post-shock term from [Lin and Eck \[2021\]](#) rather than the actual residual?

- Binary Outcome Forecasts
- Density Forecasts
- Quantile Forecasts

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