REFEREE REPORT ON OPTIMAL NON-NEGATIVE FORECAST RECONCILIATION

- 1. The abstract is too long and does not get to the point in a concise way, please reformulate.
- 2. The matrix C after eq (2) is not defined
- 3. Non-negative least square is not new can you provide references since the beginning?
- 4. Assuming that Λ_h is positive definite does not seem so straightforward, that means that forecast errors are linearly independent, but if variables are highly collinear this might not happen. I think it all depends on how you compute $\hat{y}_t(h)$. If forecast capture all comovements than I can believe the assumption. Some more explanation should be given. I see you compute univariate forecasts (via ARMA) but why not using multivariate methods in high-dimensions, e.g. factor models (see Stock and Watson, 2002), this would control better Λ_h .
- 5. "large size of the structures that typically arise in forecast reconciliation" what are the orders of magnitude? Your application has 555 series which can be handled by mulitvariate forecasting methods.
- 6. When you define λ^* that is equal to the derivative of q wrt b while you?re writing the opposite fix the notation.
- 7. Why in (9) you use the norm to define a quadratic loss? Keep the same notation as before.
- 8. Given the computational cost of running a constrained minimization with respect to the unconstrained case, what would happen if once we compute (1) we just throw away the negative forecasts and compute \tilde{y} with the remaining ones? Are we really after all m forecasts? Can you provide more motivation? And anyway what happens if you just impose the negative forecast to be zero. Can you repeat Table 7 in this case? Still the numbers there do not seem dramatically large.
- 9. What is an ETS model?
- 10. You say you log-transform data to compute forecasts and then you back-transform but doesn?t this amount to taking an exponential and therefore you must get positive values? This part should be clarified.

References

James H Stock & Mark W Watson (2002) Forecasting Using Principal Components From a Large Number of Predictors, Journal of the American Statistical Association, 97:460, 1167-1179