



The EpiBench Platform to Propel AI/ML-based Epidemic Forecasting:

A Prototype Demonstration Reaching Human Expert-level Performance

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Background

- “covid forecasting” on Google Scholar: ~76,000 results

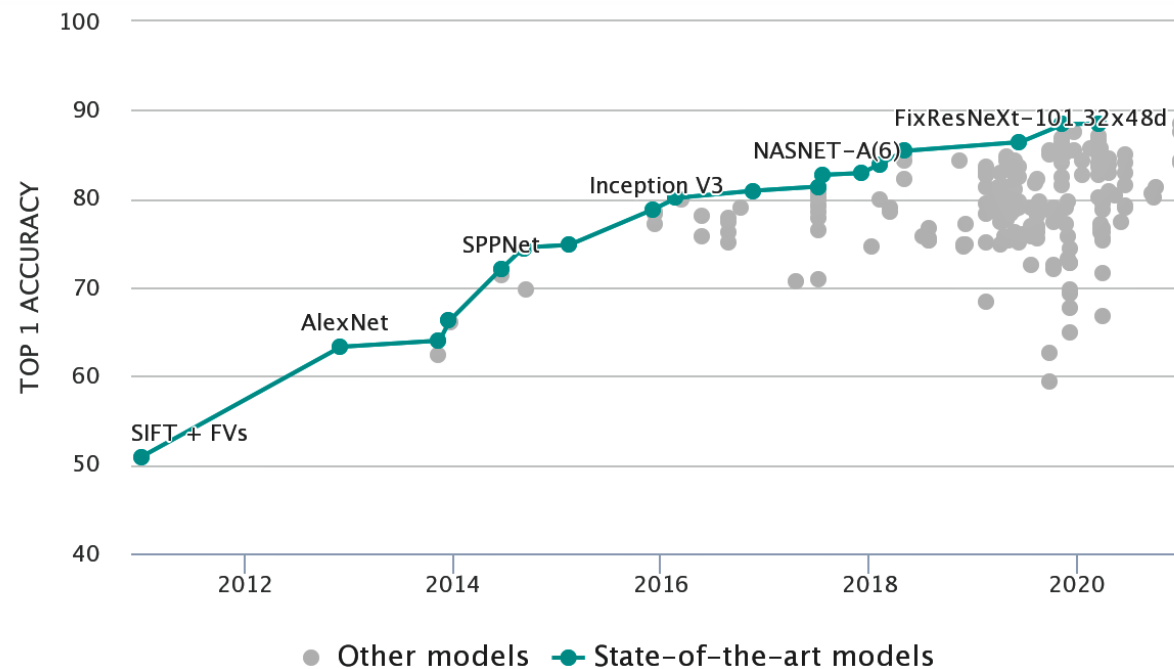


- Claims at recent AI/ML conferences: AAAI, IEEE BigData
- AI/ML claims need proper evaluation
 - “state-of-the-art”
 - Reproducibility (without human intervention)



Benchmarks Help Push Boundaries

- In other domains
 - Image Classification: ImageNet/CIFAR-10/MNIST + MobileNet/VGG/...



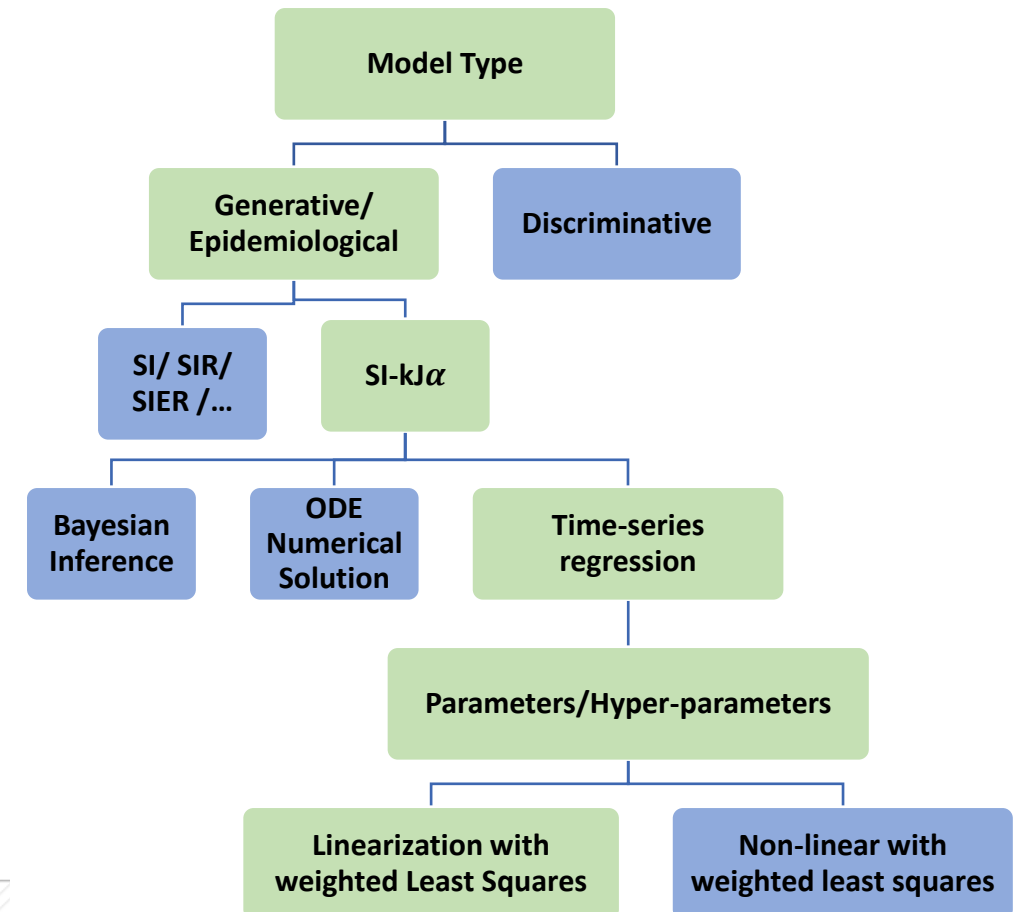
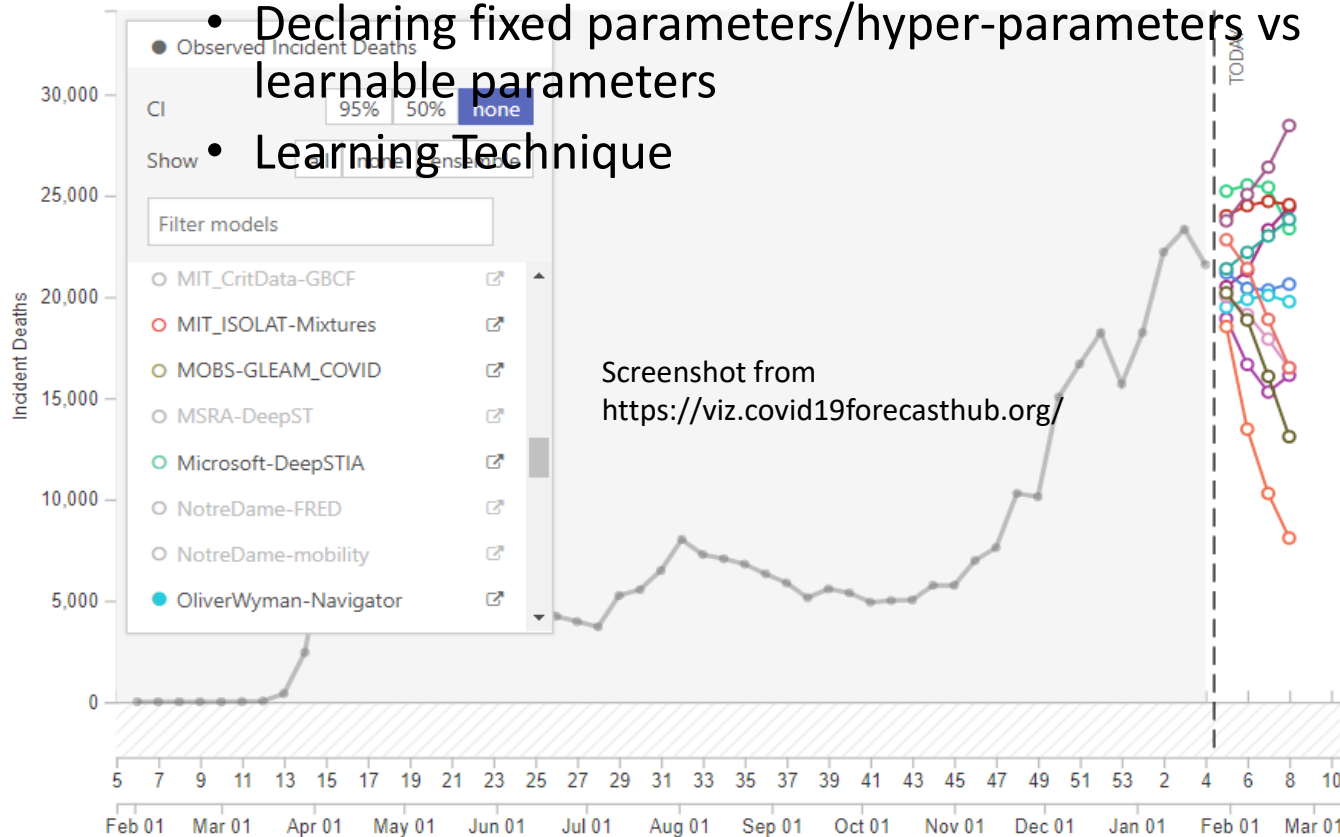


Background

"Our forecasting method outperforms SIR/SEIR"



- **Model is not everything:** The path from data to forecasts involves more
 - Data pre-processing technique
 - Declaring fixed parameters/hyper-parameters vs learnable parameters
 - Learning Technique





Existing Platforms: COVID-19 US Forecast Hub

The COVID-19 Forecast Hub

This site maintains the authoritative, up-to-date record for forecasts of COVID-19 cases, deaths and hospitalizations in the US, created by dozens of leading infectious disease modeling teams from around the globe, in coordination with the US CDC.

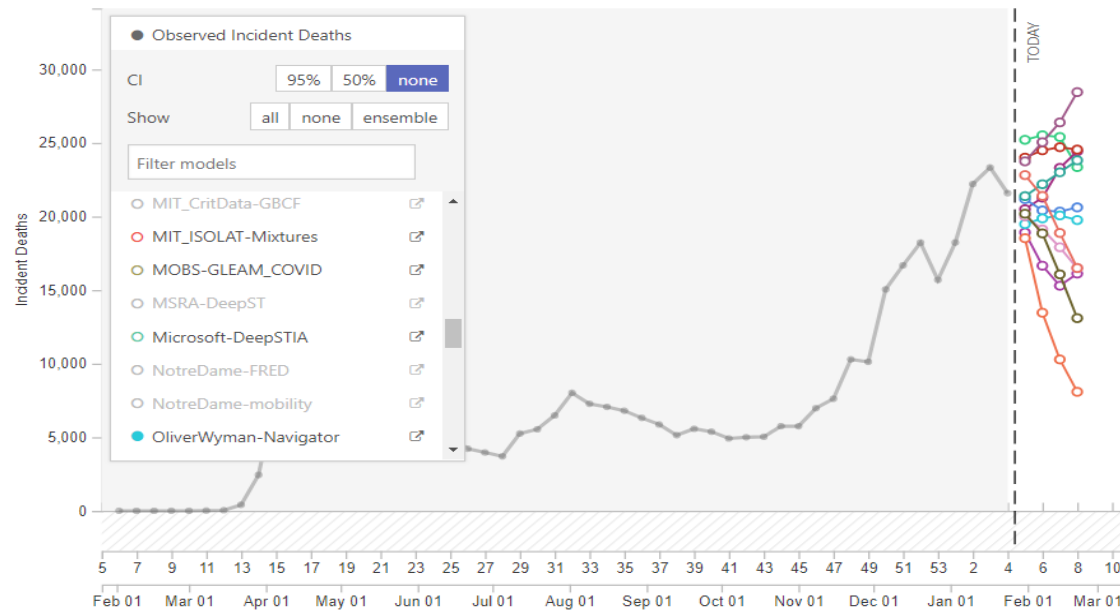
Weekly Reports

Visualization

Community

Other platforms:

- COVID-19 Germany+Poland Forecast Hub (ECDC)
- CDC Epidemic Prediction Initiative



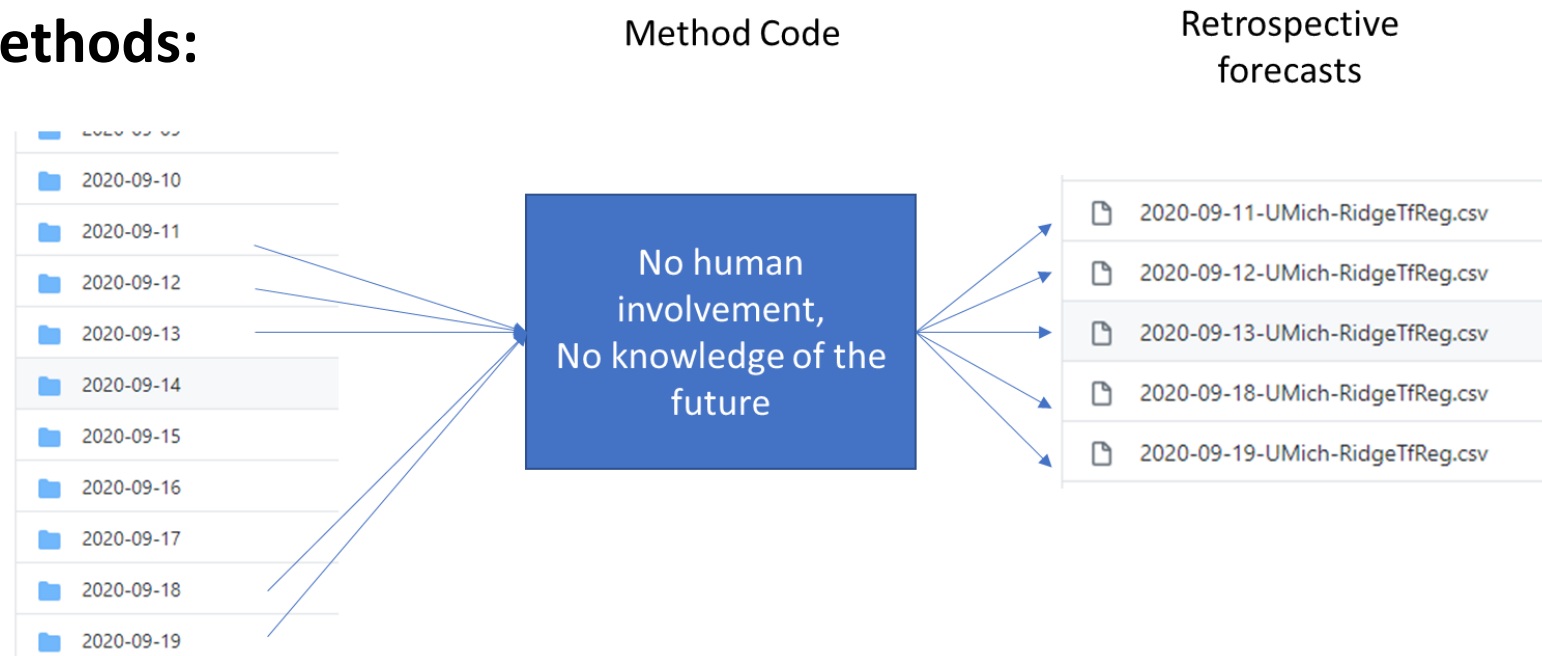
- **Forecast-hub et. al.:** Provide the best guess for the future for an ongoing epidemic
- **Forecasting benchmark:** Provide performance reference for future ML methodologies
 - MAE per region, mean MAE, ...



We are Pushing for AI/ML vs “Human” Methods

- **Human Methods:** For the ongoing epidemic forecasting, periodically “experts” update forecasts based on intuition, changing data, ...
 - Not a scalable approach to thousands of locations
 - Expertise stays with the expert

- **AI/ML methods:**





Long-term Goal

- To enable ML-driven research for epidemic forecasting by providing a collection of benchmarks (datasets + methods) that reflect the performance of various methodologies
 - For new methods to compare against
 - *Which method is state-of-the-art?*
 - *Can AI/ML with no human intervention outperform human-expert forecasts?*
- To allow a more robust ensemble learning



This work is ...

- Not claiming *“My method is the best epidemic forecasting method”*
 - But to enable such a claim
 - And to enable building better methods
- Not claiming *“I have the best evaluations”*
 - Many forecasting teams are working on evaluating ongoing COVID-19 forecasts, including us in a different project
 - How to properly evaluate is still an open problem – We hope the platform will enable this discussion too



Contributions

- Introducing a prototype: **COVID-19 Forecasting Benchmark**
 - A platform for the task of COVID-19 case and death forecasts
 - Already accepting submissions
- Demonstrate that the platform can enable fully automated (no human intervention) ensemble reaching human-expert-level performance



Comparison for New Methods

Filter: All Methods ML/AI Methods Human-Expert Methods

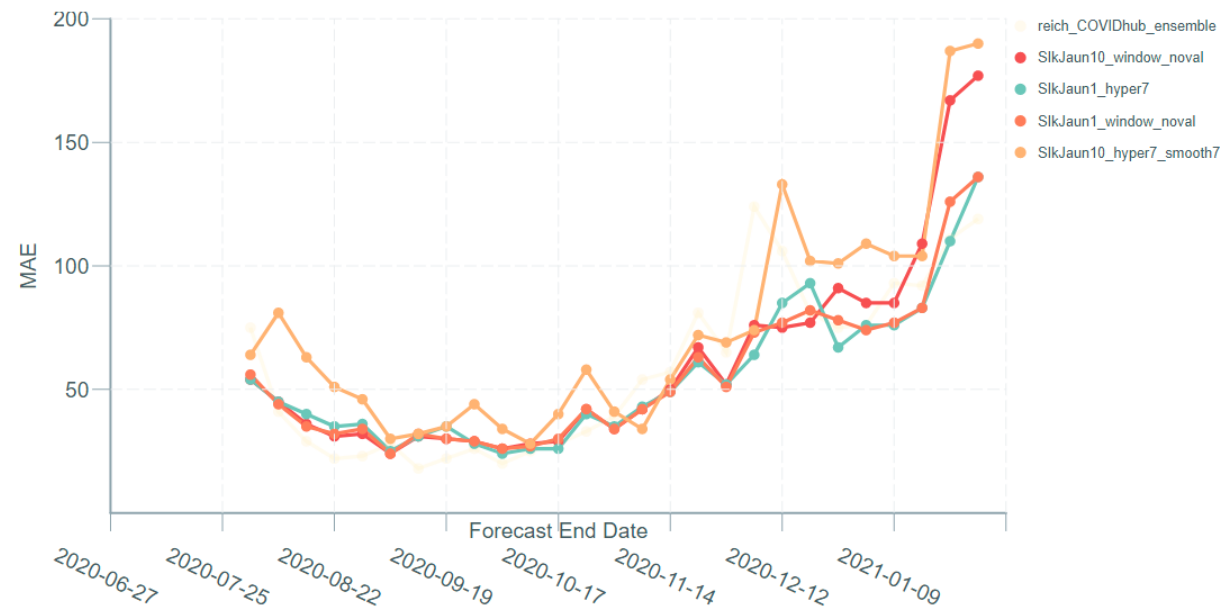
Region: US Average

Methods: reich_COVIDhub_ensemble SikJaun10_window_noval SikJaun1_hyper7 SikJaun1_window_noval SikJaun10_hyper7_smooth7

Metrics: MAE

Forecast Type: COVID-19 death US state-level death forecasts

Prediction Time Span: 1-week-ahead 2-week-ahead 3-week-ahead 4-week-ahead

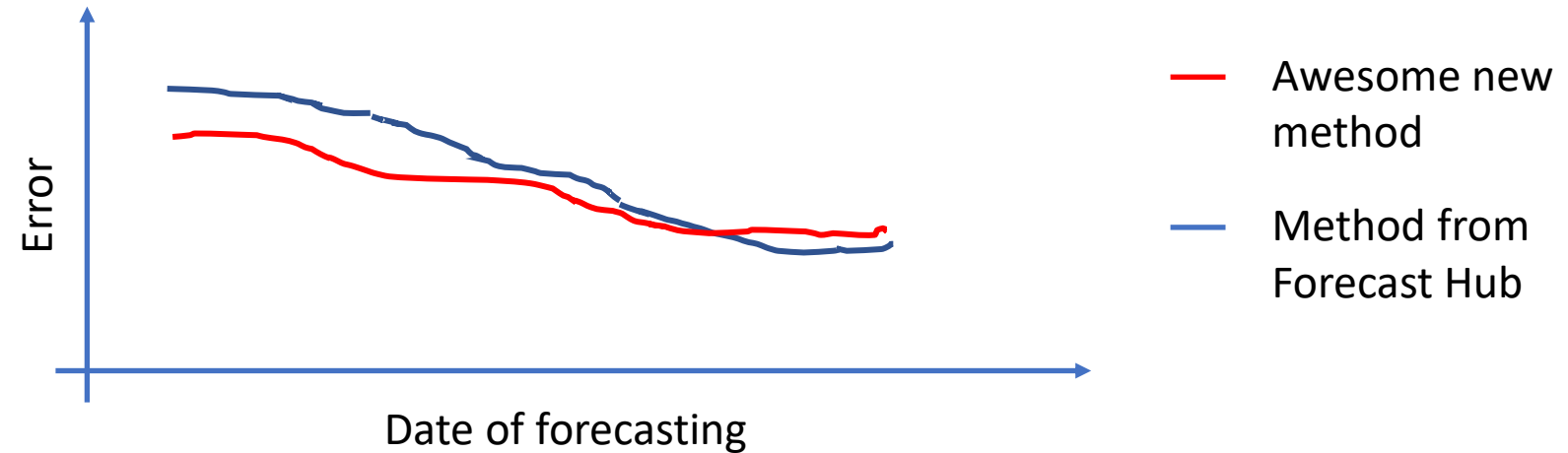


Our platform enables submissions of forecasts and comparisons of methods in an interactive way



Comparison for New Methods

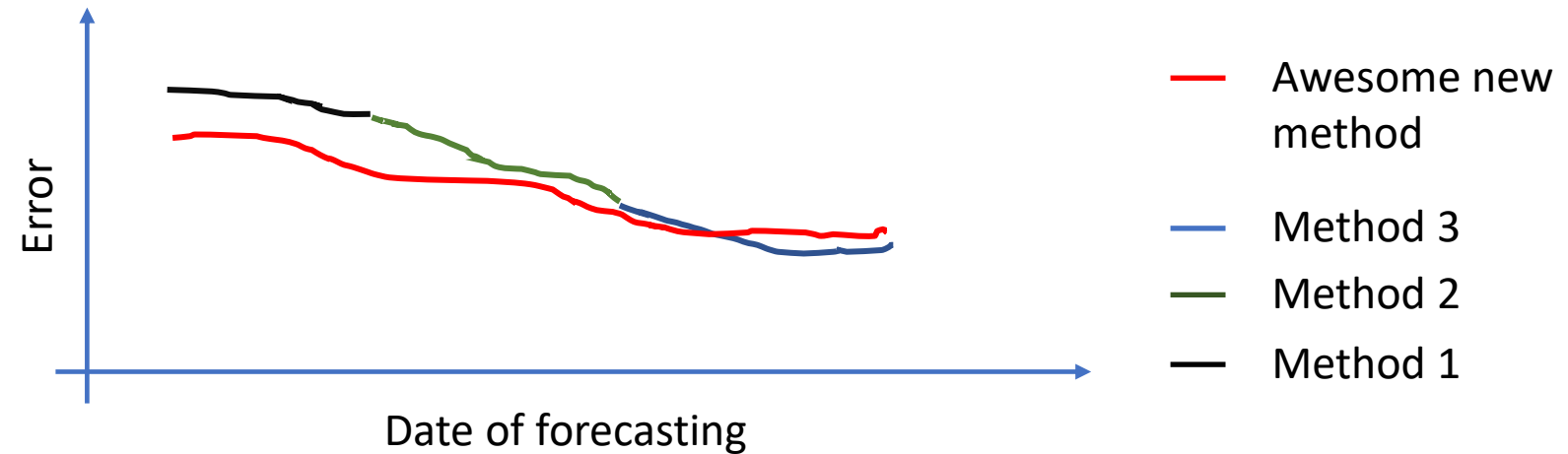
- Comparison against Forecast-hub?





Comparison for New Methods

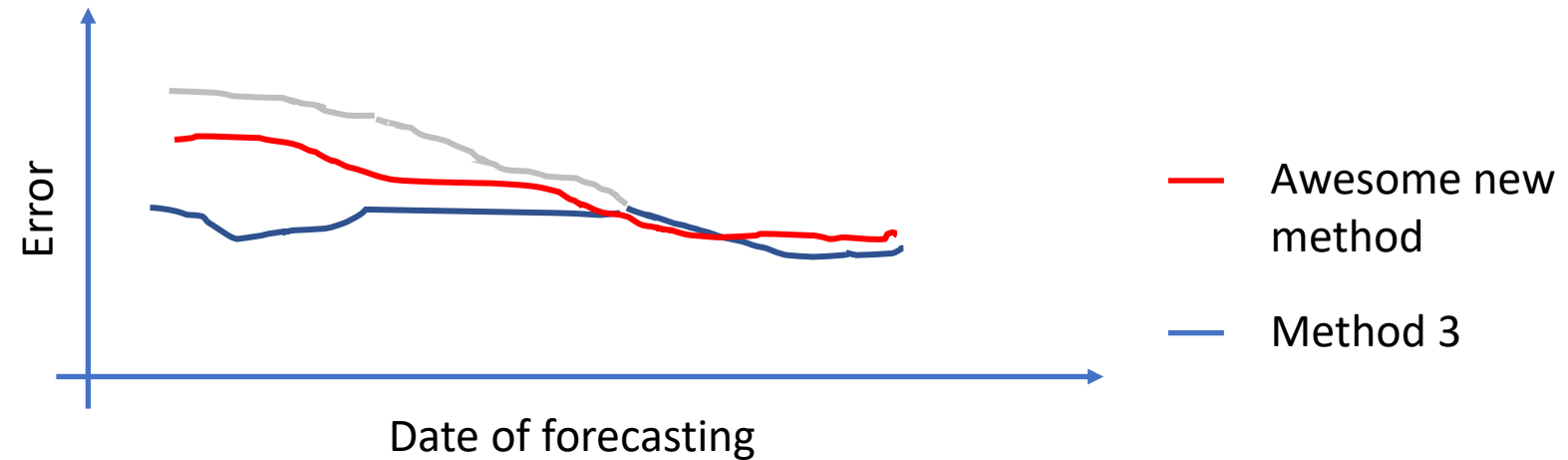
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Comparison for New Methods

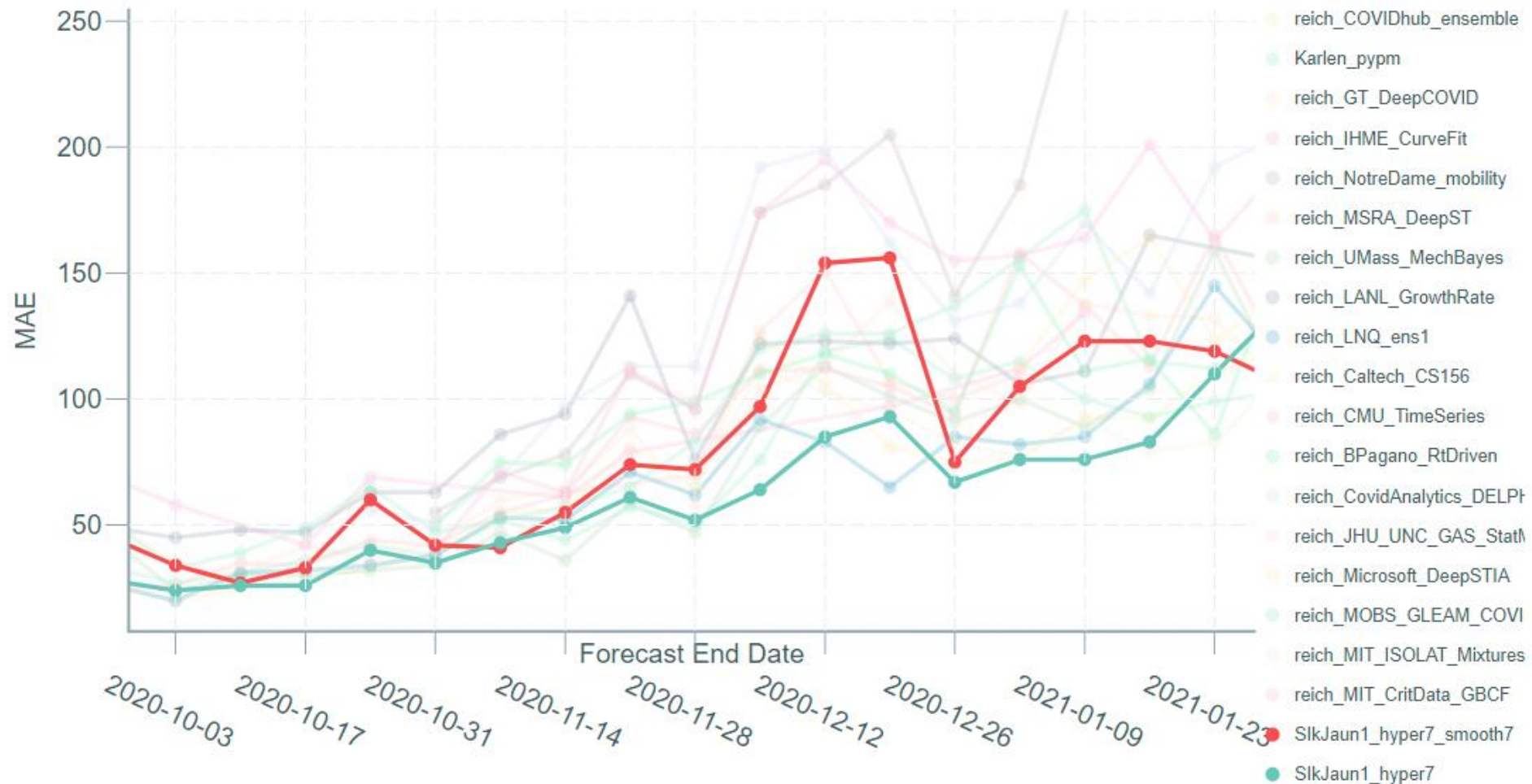
- Comparison against Forecast-hub?



“Which team is better” vs “Which method is better”



Tuning => Different method!





Ensemble Learning

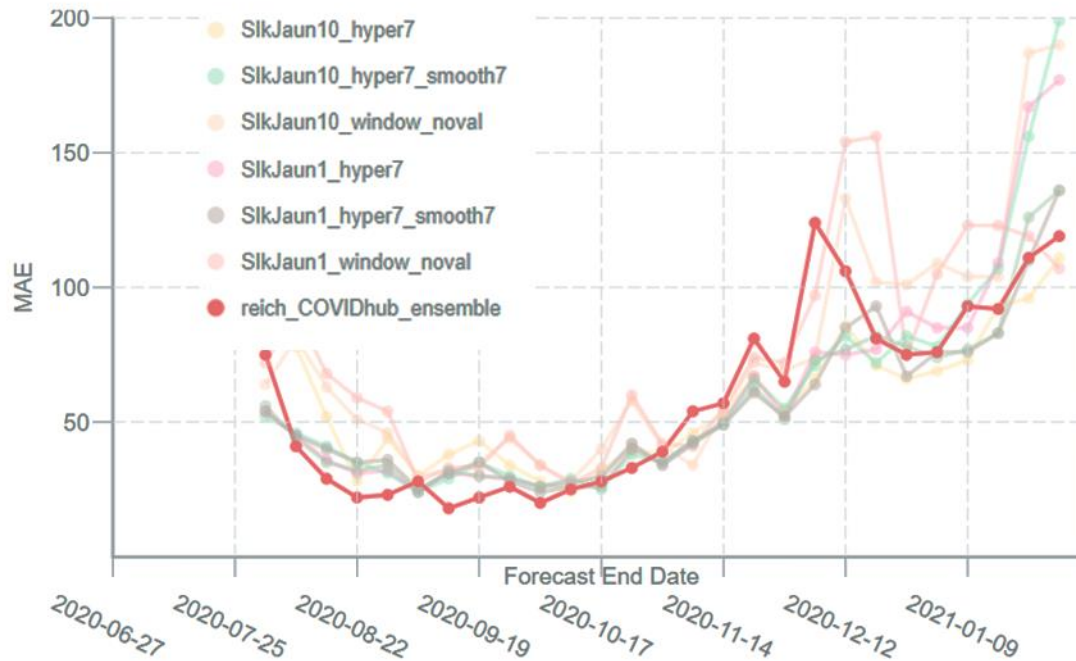
- Current ensembles still rely on simple averaging/median of individual method (constituent) forecasts
 - Among evolving methodologies, a smarter rule is difficult to learn
- With **consistent** methodologies, we can learn hidden rules on how to combine the results

- Example Approach:





Ensemble Results: 4 week ahead incident death forecasts

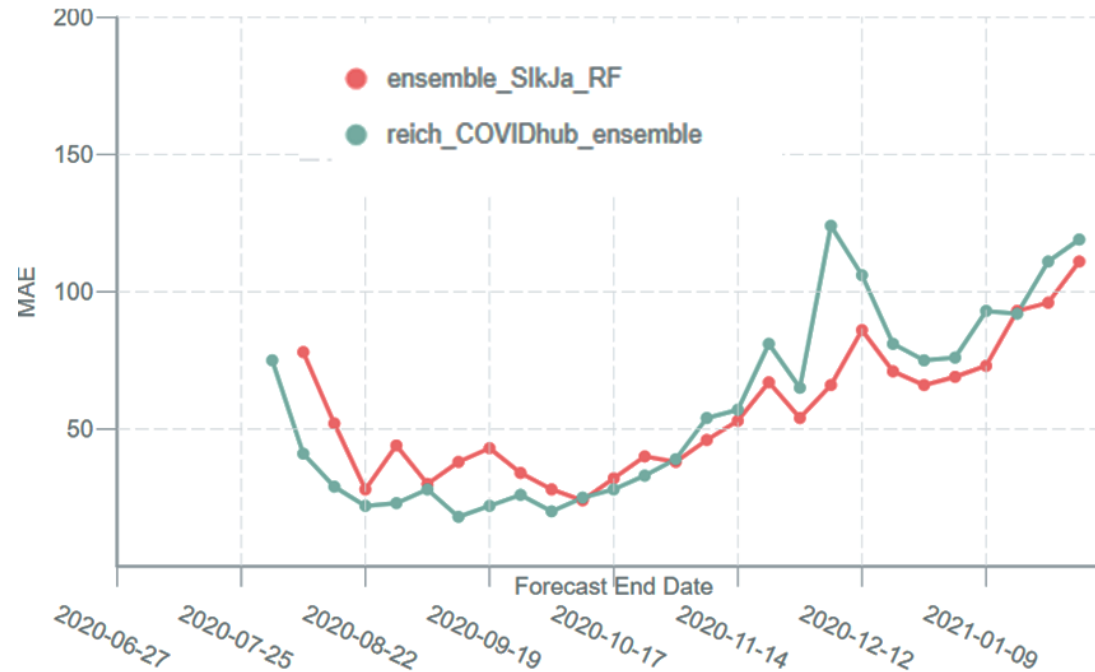


- Constituent methods: Variations of “SikJalpha” :
 - Various settings for pre-processing (e.g., smoothing factor)
 - With or without “validation set”

- **Observation:** Different constituent methods better or worse at different points in time compared to “COVIDhub Ensemble”



Ensemble Results: 4 week ahead incident death forecasts



- **Observation:** Random Forest Ensemble close to COVIDhub-ensemble, in many cases better!
- Recall:
 - **COVIDhub Ensemble:** Combination of forecasts from > 30 expert teams each tuning their methods over time
 - **This Ensemble:** Combination of fast forecasting methods, all *without human intervention*



Submissions

- Help us create this resource for AI/ML community!
- Details at: <https://scc-usc.github.io/covid19-forecast-bench>
- Thanks to those who have already submitted and/or provided details of methodology!