### VN1 Forecasting Challenge

5th Place Submission

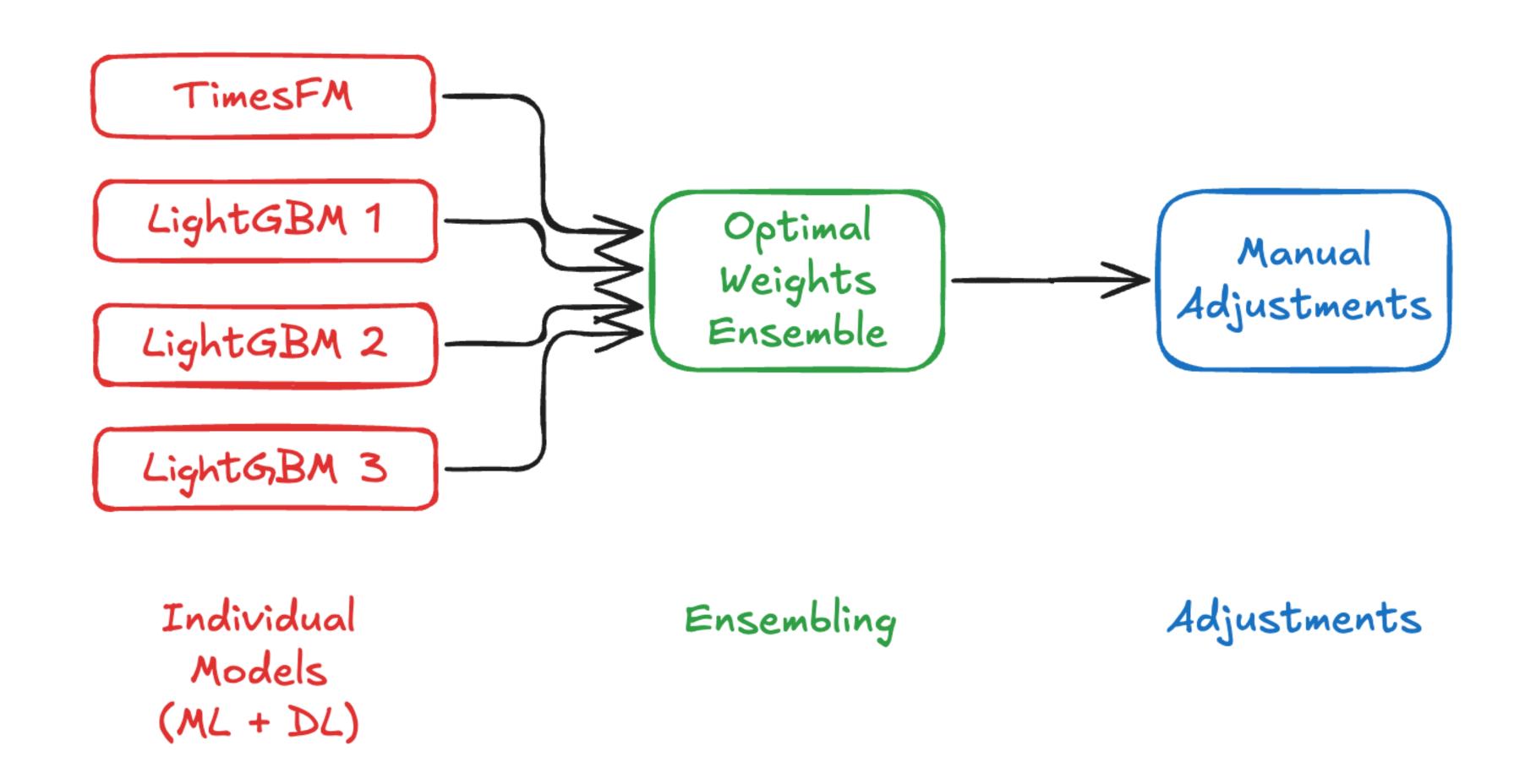
### Hey, my name is An

#### Introduction

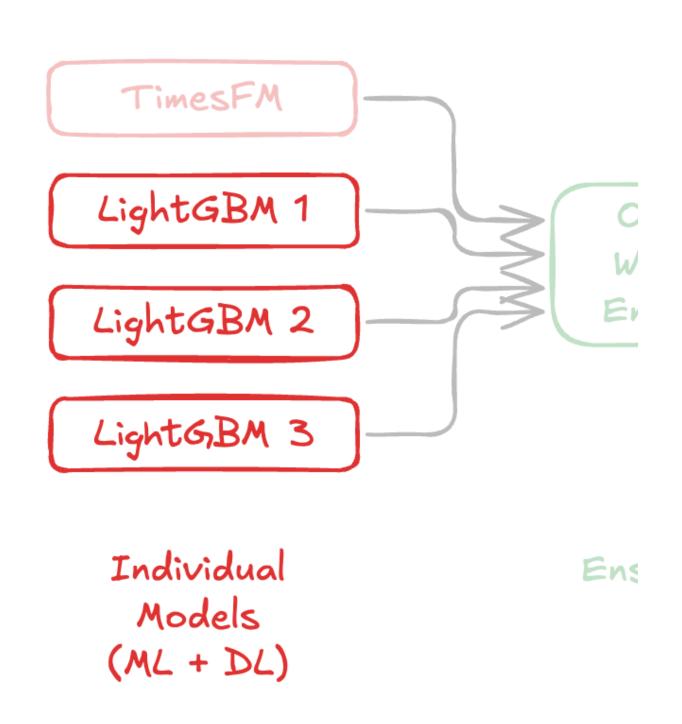
- M.Sc. in Mathematics
- Data Scientist @ dm-drogerie markt
- Amateur Chef
- 👴 1st time participating in forecasting competition

#### The forecast is an ensemble of diverse models

#### **Modelling Overview**



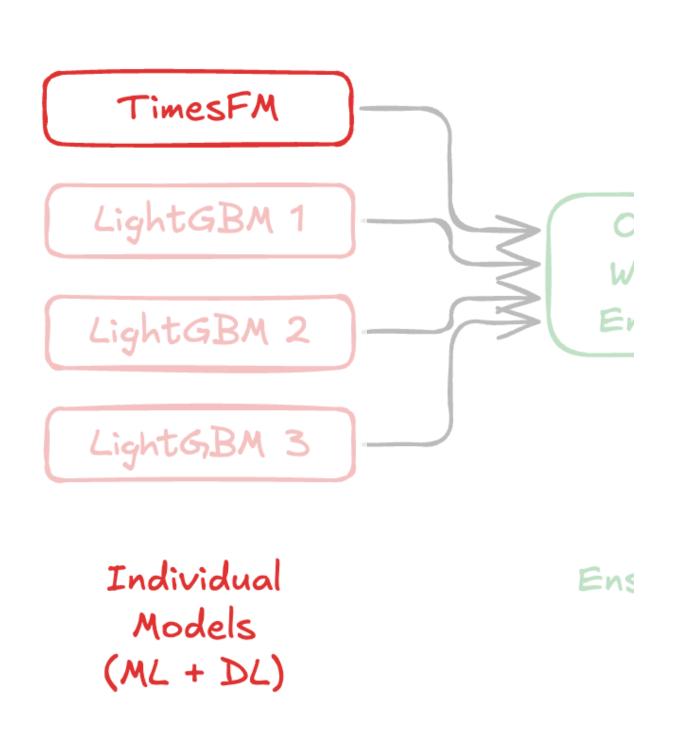
# Three recursive LGBM models were trained Individual Model (LGBM)



- Trained LightGBM Models with MLForecast
- Recursive Forecasting strategy
- Min-Max Scaling applied to target variable
- Three models: L2 / L1 / Tweedie loss functions
- Feature Engineering:
  - Date: Month, Fourier Features
  - Target: Lags, Rolling Min/Max/Mean/Std/Quantiles

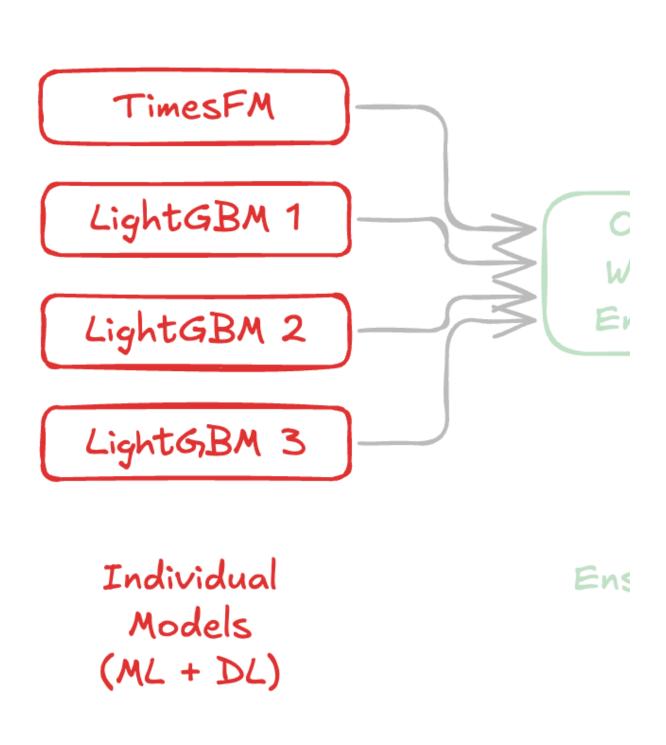
### A pretrained foundation model was included

#### Individual Model (TimesFM)

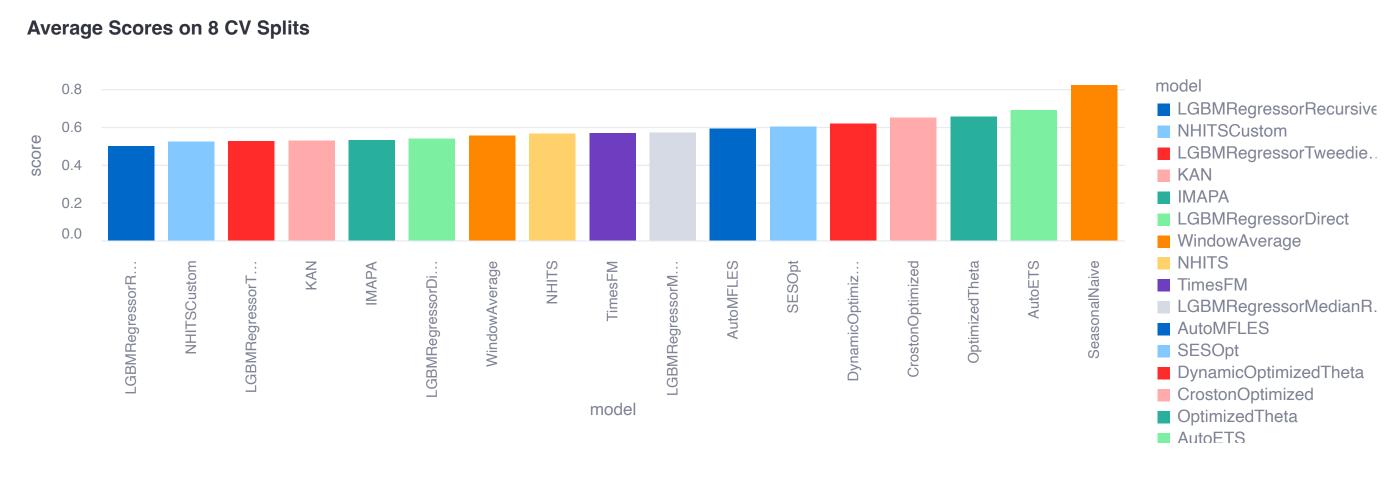


- TimesFM is a foundation model for time series developed by Google
- The checkpoint google/timesfm-1.0-200m-pytorch was used to generate the predictions

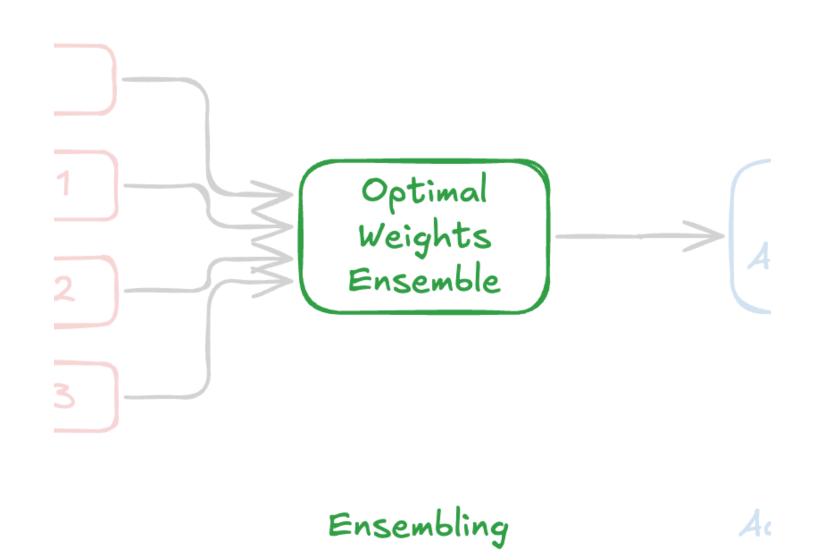
# The models were chosen from a large collection Individual Model



- Many models from deep learning, statistics, and machine learning were tested, but not all were included in the final ensemble
- The final models were selected for their performance and stability across multiple cross-validation splits

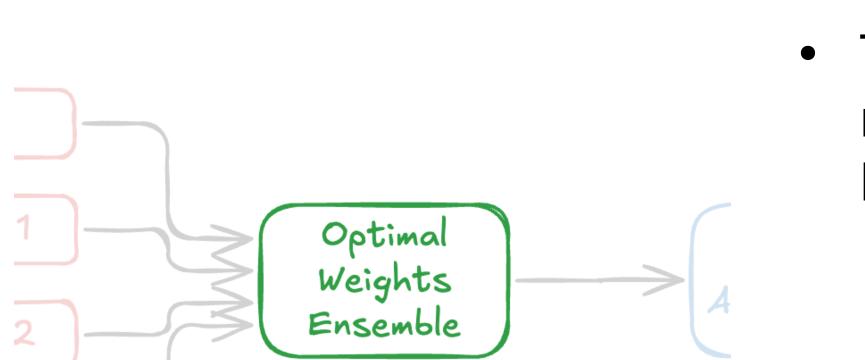


# The models are combined by optimizing weights Optional Weights Ensemble



- Goal: Combine predictions of different individual models to achieve the most accurate overall forecast.
- Weighted Models: Assign a weight to each model's prediction, then combine them to create a single forecast.
- Minimizing Error: Choose the weights by minimizing the toal total error (0.5 \* Bias + 0.5 \* MAE).

# The top model isn't always the most weighted Optional Weights Ensemble



Ensembling

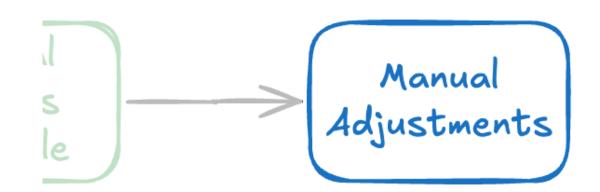
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 The model with the highest score is not necessarily the model with the strongest individual performance in the backtesting



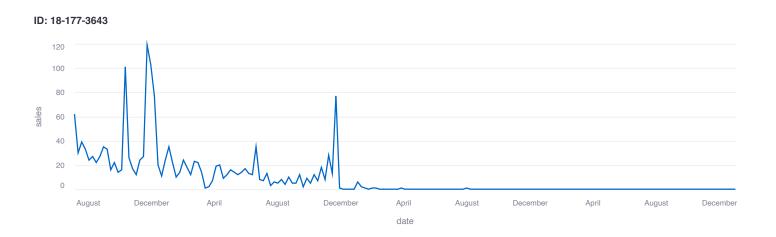
### The forecast is adjusted manually afterwards

#### Manual Adjustments

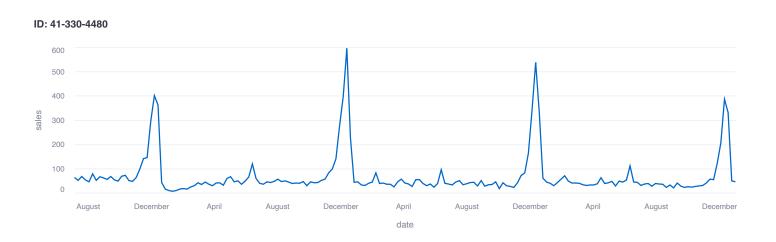




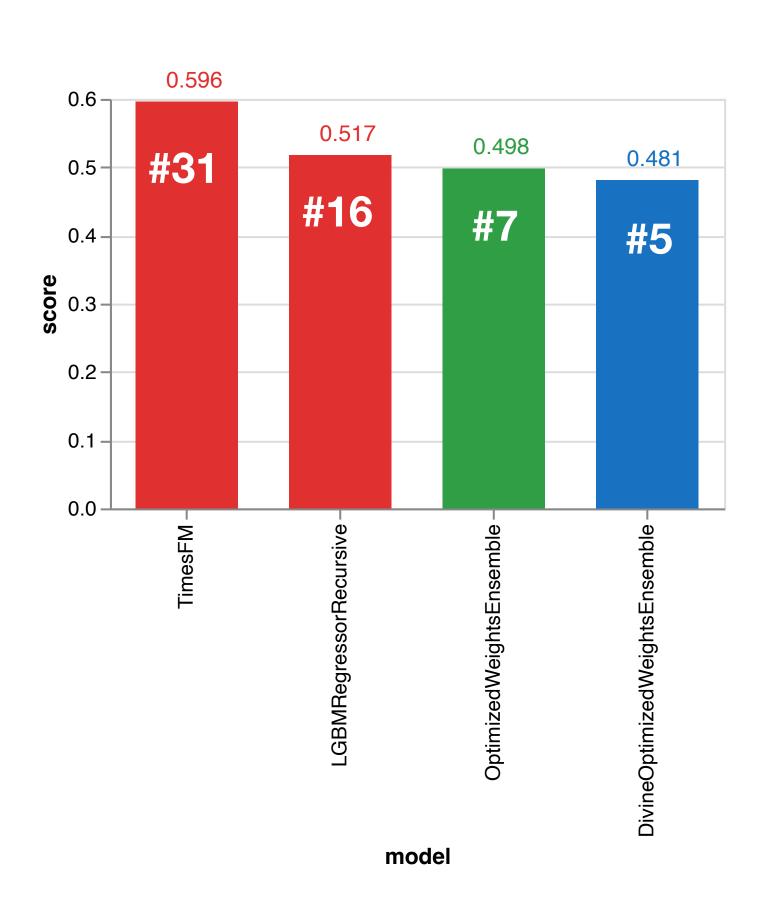
 Zeros: Time series with no sales in the last 13 days are forecasted with constant zeros



• Seasonal: Time series with a high seasonal ACF score are forecasted using a seasonal naive model



# **Each modeling step improved the final model**Final Result



- The individual models alone wouldn't have been able to get into the top five
- Surprisingly, foundation models did improve the final ensembling, even if they alone perform worse than simple benchmark models
- Manual intervention can proactively prevent the model from making significant mistakes and improve the score

# Thank you for your attention $\checkmark$