## Method \& <br> Process



Crisp \& Green is a fast-casual, healthy salad and grain bowl restaurant with 8 locations across MN and TX

Labor costs represent one of the food-service industries' largest variables and most significant hurdles. Poor labor efficiency can handicap a restaurant's profits

The current staffing approach had been to use a time-intensive simple moving average model, resulting in inefficient labor supply across non-peak hours, and thus excess labor cost

Our project seeks to analyze customer and sales data to project sales by 15 -minute intervals to create better, more efficient labor staffing models and reduce excess cost

## Method \&

## Process



Our solution utilizes sales data out of the client's NCR system, which is then cleaned to remove erroneous values and better fit targeted sales demand

We employed 6 different Machine Learning Regression \& Time Series models (OLSR, KNN, Random Forest, SVM, Multivariate Moving Average, and ARIMA) to find the best method for sales predictions

| Date | Tender_Hour | Tender_Minute | Tender_Amount | Tender_Type_Name |
| :---: | :---: | :---: | :---: | :---: |
| Date | 16 | 12 | \$ | *Hidden* |
| Date | 13 | 0 | \$ | *Hidden* |
| Date | 10 | 32 | \$ | *Hidden* |
| Date | 10 | 36 | \$ | *Hidden* |
| Date | 18 | 18 | \$ | *Hidden* |
| Date | 18 | 20 | \$ | *Hidden* |
| Date | 18 | 24 | \$ | *Hidden* |
| Date | 20 | 21 | \$ | *Hidden* |
| Date | 10 | 36 | \$ | *Hidden* |
| Date | 2 | 0 | \$ | *Hidden* |
| Date | 9 | 34 | \$ | *Hidden* |
| Date | 13 | 11 | \$ | *Hidden* |
| Date | 14 | 18 | \$ | *Hidden* |
| Date | 15 | 49 | \$ | *Hidden* |
| Date | 10 | 4 | \$ | *Hidden* |
| Date | 11 | 23 | \$ | *Hidden* |
| Date | 15 | 3 | \$ | *Hidden* |
| Date | 9 | 37 | \$ | *Hidden* |
| Date | 9 | 42 | \$ | *Hidden* |
| Date | 14 | 50 | \$ | *Hidden* |
| Date | 16 | 3 | \$ | *Hidden* |

## Results \&

Conclusion


The Stacking Ensemble method had the highest average \% improvement in RMSE over the client's current prediction method (avg of $6.5 \%$ for a given week)

We chose the multivariate moving average model (where the weights were determined using a hyperparameter grid search built via $R$ ) due to the ease of upkeep and automation

The final program automatically compiles usable weekly reports also visible at granular levels of 15 segments minutes per day

Thursday 04-29-21 and Thursday 05-06-21


Friday 04-30-21 and Friday 05-07-21


Saturday 05-01-21 and Saturday 05-08-21


