AI POWERED FOOD DEMAND FORECASTING

LITERATURE SURVEY

YEAR	AUTHOR	OBJECTIVE	METHODOLOGY	LIMITATION
2020	K.Aishwarya Aishwarya.N.Rao Nikita Kumari, Akshit Mishra, Mrs.Rashmi M R	It provide an appropriate algorithm for demand forecasting which is capable of overpowering the wastage of short life items.	The number of customers is forecasted using machine learning and statistical analysis methods with internal data and external data in the ubiquitous environment. Bayesian Linear Regression, Boosted Decision Tree Regression, and Decision Forest Regression are used for machine learning, Stepwise method is used for statistical analysis methods. Their by predict it.	This evaluation is used practically for restaurants. More refined prediction can be done based on many other factors like cultural habits, religious holiday, consumer preferences.

To validate the 2022 To gain more we estimated Vera LuciaMigueis the demand for information about proposed fresh fish in a their customers and approach and AndrePereira representative their buying behavior. create a JoaoPereira This data collection comprehensive store of a large GoncaloFigueira European has promoted a huge assessment of retailing opportunity for predictive models company used improving operations. for fish demand Thus, retailers have as a case study. forecasting, we The results concentrated on have revealed that the implemented and developing more tuned some of the machine accurate forecasting learning models best-known models that help them make decisions techniques from provided accurate that are more the literature. data-driven and less forecasts in Furthermore, two intuition-based baseline models comparison to the baseline were constructed. models and the One considered statistical as demand model, with the estimates the demand observed Long **Short-Term** on the **Memory** homologous day networks model of the previous yielding, in week and the general, the best other the demand results in terms observed on the day when the of root mean squared error estimates were (FESS). computed.

2018 Bohdan M. the differentials Deep learning models Often traditional Pavlyshenko of the predictive demand forecasting have approach in been tested and can create an relation to confirmed for overlap between demand, taking forecasting crude oil correlation of into account demand patterns prices, photovoltaic power and the causes of factors such as and on-demand ride seasonality, fluctuations in geographic / services. demand. At times, Research results related regional food retailers may preferences and to the food industry use techniques to changes in mention deep learning extract customer methods consumer behavior patterns behaviour. (Convolutional Neural from correlations in Network (CNN)-based demand shifts and food image link to external recognition algorithm) events. This may used to derive food cause them to think information (food that this is linked to type and portion size) demand shifts and from food image or to believe it is the true propose an assistive cause of it. calorie measurement system. In proposed a time-dependent food distribution model and a Weight optimisation algorithm aimed at adapting the user's data to their eating habits. Deep learning has also been imposed in the waste sorting process to automate some of the waste handling tasks.

2018 Takashi Research has been There was no big The number of Tanizakia, conducted on how difference in the customers is forecasted Tomohiro to advance store forecasting rate using machine learning Hoshinoa, using the method of management by and statistical analysis Takeshi improving Bayesian, Decision, method with internal employees' work Shimmurab, and Stepwise, and data and external data in Takeshi arrangement and the forecasting rate the ubiquitous food materials Takenakac of Boosted was a environment. Bayesian ordering based on little low. The Linear Regression, accurate forecast rate of any **Boosted Decision Tree** forecasting of the store exceeded Regression, and number of approximately **Decision Forest** customers for 85%. We got the Regression are used for face-to-face evaluation that this machine learning, service industries method is Stepwise method is and demand practically used for statistical forecasting applicable from the analysis method. methods restaurant. conducted by using internal data such as POS data and external data in the ubiquitous environment such as weather, events, etc. in order to improve the accuracy of demand forecasting

Patrick Meulstee, Mykola Pechenizkiy An ensemble learning approach is employed for dynamic integration of classifiers for better handling of seasonal changes and fluctuations in consumer demands. It focuses on currently operated, and how to improve predictions for each product by constructing new groups of predictive features from (1) publicly available data about the weather and holidays, and (2) data from related products. An ensemble learning approach on the real data collected by food wholesaling and retailing company. The results demonstrate that our ensemble learning approach can perform much better than the currently used baseline, we can handle seasonal changes with our ensemble learning approach can perform much better than the currently used baseline, we can handle seasonal changes with ensemble learning better if feature set for a target product is complemented with features of related product, and an ensemble can become more accurate if information about the weather and holidays is presented explicitly in a feature set. we studied this idea only with ensemble learning approach and implicit use of information about the changes observed in related products in base classifiers	-				
	2008	Mykola	learning approach is employed for dynamic integration of classifiers for better handling of seasonal changes and fluctuations in consumer demands. It focuses on currently operated, and how to improve predictions for each product by constructing new groups of predictive features from (1) publicly available data about the weather and holidays, and (2) data from	approach on the real data collected by food wholesaling and retailing company. The results demonstrate that our ensemble learning approach can perform much better than the currently used baseline, we can handle seasonal changes with ensemble learning better if feature set for a target product is complemented with features of related product, and an ensemble can become more accurate if information about the weather and holidays is presented explicitly in a	showed that we can handle seasonal changes with ensemble learning for many products better if feature set for a target product is complemented with features of the most related products, and that an ensemble can become more accurate if information about the weather and holidays is presented explicitly in a feature setwe studied this idea only with ensemble learning approach and implicit use of information about the changes observed in related products in base