# LightGBM Model For Predict Future Sales

**Kaggle Competition** 

#### **Table of Contents**

- Data Pre-processing
- Feature Engineering
- Model
- Data Post-processing

# **Data Pre-processing**

Applying EDA (Exploratory Data Analysis).

# **Feature Engineering**

- Mean encoding
- Lag features

# Mean encoding

Calculate mean for groupby features.

# Example:

mean(groupby(shop\_id))

→ return number of products/month of that shop

# Lag features

Use **past** values as features for predicting **future** values.

# **Model - LightGBM**

A gradient boosting framework that uses tree-based learning algorithms

# Model - LightGBM's key features

- Optimization in Speed and Memory Usage
- Optimization in Accuracy
- Support for parallel and GPU learning
- Optimal for categorical features
- Handle large-scale data

# Model - How LightGBM's work

- Gradient Boosting: uses gradient boosting framework.
- Decision Trees: Builds decision trees one at a time, where each new tree corrects errors made by the previously trained tree.
- Leaf-wise Growth: Grows tree leaf-wise rather than level-wise

# **Model - Hyperparameters**

- Learning rate
- Number of trees
- Max depth
- Feature fraction
- Bagging fraction

# **Data post-processing**

#### Reference:

https://www.kaggle.com/code/abubakar624/first-place-solution-kaggle-predict-future-sales