





By: Ilham Rofi'i, N.







Overview

Objective 1	Memprediksi pembelian produk
Objective 2	Seperti apa produk yang laris ?









Dataset

Open CDP

Oktober 2019 – Februari 2020

event_time

waktu tiap terjadinya event/perilaku dalam UTC

event_type

perilaku yang dilakukan, ada 3 jenis yaitu view, cart, dan purchase product_id

ID dari produk

category_id

ID untuk kategori produk

category_code

Kategori produk

brand

nama brand produk

price

harga produk

user_id

ID permanen pengguna

user_session

4412

ID pengguna tiap yang berubah tiap sesi



Exploratory Data Analysis

Analisis kolom category_code level dasar

Analisis kolom category_code lengkap

Analisis kolom brand

Analisis brand produk top 10 kategori produk

Analisis kolom price

Analisis price produk top 10 kategori produk

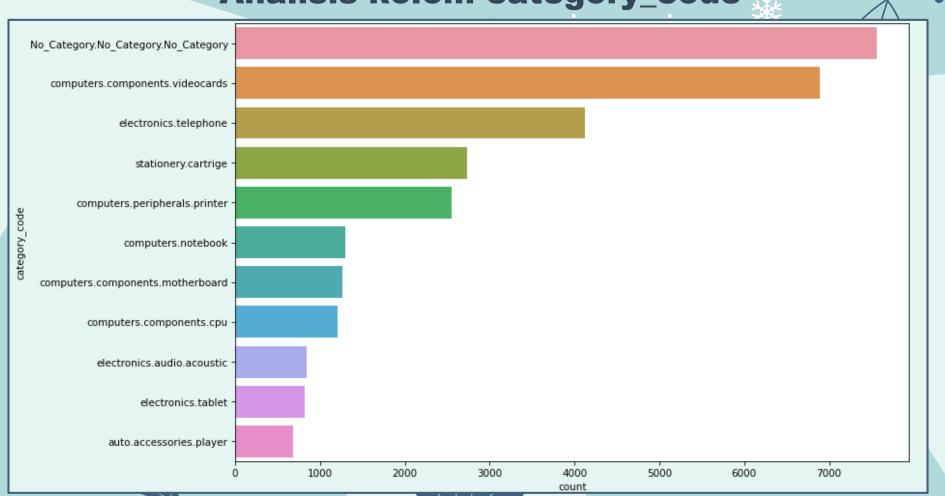
Analisis kolom activity

Analisis kolom weekday



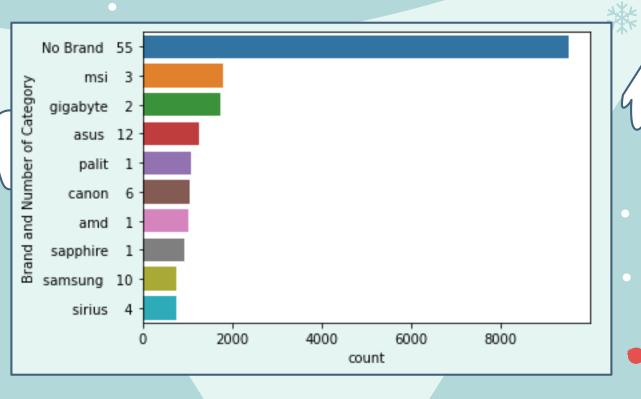


Analisis kolom category_code





Analisis kolom brand









Analisis brand produk top 10 kategori produk

	Top 3 Brand	Persentase No Brand
videocards	gigabyte, No Brand, sapphire	10%
telephone	No Brand, sirius, No Brand	61%
cartrige	hp, canon, nv-print	33%
printer	canon, pantum, canon	22%
notebook	mobilepc, No Brand, topon	28%
motherboard	asrock, asus, asrock	0%
сри	amd, amd, amd	0%
acoustic audio	edifer, edifer, edifer	4%
tablet	samsung, xiaomi, irbis	35%
player (auto accessories)	pioneer, kenwood, pioneer	0%











Analisis price produk top 10 kategori produk







	product_id	category_code	brand	price	count	cheaper_percentage
0	4099645	computers.components.videocards	gigabyte	292.83	564	22.24%
1	1821813	computers.components.videocards	No Brand	397.48	538	51.04%
2	893196	computers.components.videocards	sapphire	214.10	384	8.92%
0	3829355	electronics.telephone	No Brand	32.22	543	38.44%
1	1821557	electronics.telephone	sirius	16.19	121	6.23%
2	3632668	electronics.telephone	No Brand	34.44	116	82.81%
0	3642540	stationery.cartrige	hp	26.19	345	21.91%
1	1785245	stationery.cartrige	canon	43.11	207	69.79%
2	471387	stationery.cartrige	nv-print	13.86	110	0.00%
0	799067	computers.peripherals.printer	canon	82.97	233	8.69%
1	3829912	computers.peripherals.printer	pantum	122.86	122	41.58%
2	3790736	computers.peripherals.printer	canon	163.52	94	55.01%
0	1044616	computers.notebook	mobilepc	22.22	95	9.29%
1	3627467	computers.notebook	No Brand	31.11	46	45.36%
2	1006967	computers.notebook	topon	55.87	46	87.43%





Analisis price produk top 10 kategori produk







	product_id	category_code	brand	price	count	cheaper_percentage
0	136700	computers.components.motherboard	asrock	77.73	194	69.05%
1	809948	computers.components.motherboard	asus	64.48	187	26.73%
2	523117	computers.components.motherboard	asrock	73.81	125	54.40%
0	3791351	computers.components.cpu	amd	204.29	423	40.22%
1	942339	computers.components.cpu	amd	56.27	96	0.00%
2	3804514	computers.components.cpu	amd	140.86	79	27.82%
0	1271550	electronics.audio.acoustic	edifier	84.89	76	14.97%
1	124712	electronics.audio.acoustic	edifier	204.79	61	67.07%
2	471287	electronics.audio.acoustic	edifier	117.75	43	48.50%
0	3829572	electronics.tablet	samsung	57.62	67	30.58%
1	1578612	electronics.tablet	xiaomi	284.30	58	82.26%
2	1804316	electronics.tablet	irbis	127.08	40	70.03%
0	4099764	auto.accessories.player	pioneer	126.68	180	36.09%
1	4100427	auto.accessories.player	kenwood	104.86	68	11.09%
2	1843507	auto.accessories.player	pioneer	153.83	50	84.57%







Analisis kolom activity

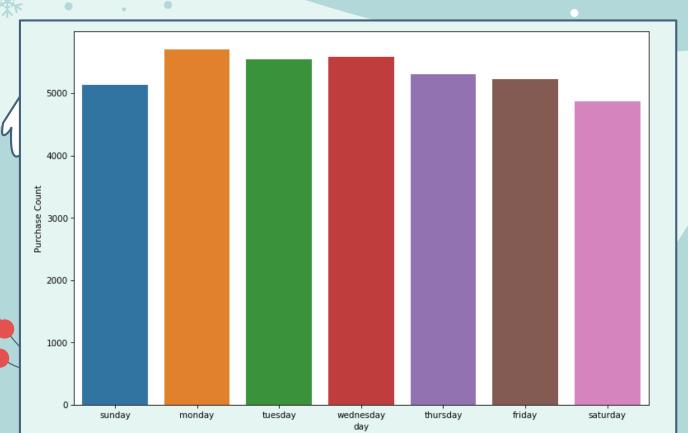








Analisis kolom weekday



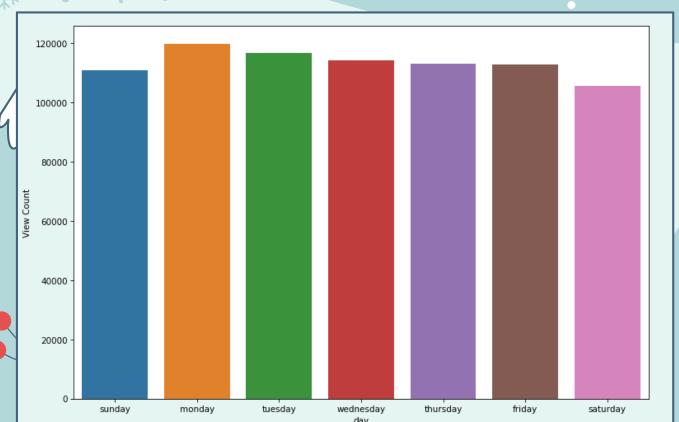








Analisis kolom weekday









Analisis Machine Learning

Objective 1	Apakah feature dalam dataset mempengaruhi target variable ?
Objective 2	Feature apa yang paling penting ?
Data Preparation	feature engineering, data cleansing, overcoming imbalances, categorical variable encoding
Data Modeling	Xgboost Machine Learning Algorithm
Feature Importance	xgboost built-in feature, permutation methode, SHAP







Data Preparation

feature engineering	'brand', 'price', 'event_weekday', 'category_code_level1', 'category_code_level2', 'activity_count'
data cleansing	Drop duplicated data Fill missing data
encoding	Using LabelEncoder() from sklearn
data imbalances	45% Imbalances







F1-Score

0.63

Data Modeling











Feature Importance



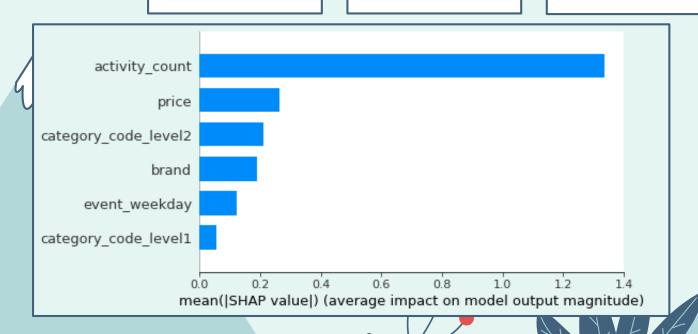
SHAP

Metode feature importance utama

Permutaion Methode

Sama persis dengan hasil SHAP XGBoost Built-in Function

Sedikit berbeda dengan hasil SHAP



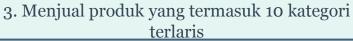


Rekomendasi



1. Menjual barang yang populer

a. Barang dengan brand terkenal b. Melakukan kegiatan pengiklanan



videocards, telephone, cartrige, printer, notebook, motherboard, cpu, acoustic audio, tablet, player (auto accessories)

2. Menjual produk dengan harga murah hingga mahal

Produk paling laris harganya dibawah 50% barang terjual di kategori yang sama.

4. Menjual produk dengan brand terkenal

msi, gigabyte, asus, palit, canon, amd, sapphire, samsung, sirius











Terima Kasih atas Perhatiannya









slidesgo