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boxcoxlp\npd.set_option('display.float_format', lambda x:
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81)\npd.set_option('display.max_rows',
101)\npd.set_option('display.max_colwidth', 100)\n\n'''Seaborn and
Matplotlib Visualization'''\nimport matplotlib.pyplot as plt\nimport
seaborn as sns\nplt.style.use('bmh')
\nsns.set_style({'axes.grid':False})
\nsns.set_style('whitegrid')\n%matplotlib
inline\n\n'''Validation'''\nfrom sklearn.model_selection import KFold,
cross_val_score\n\n'''Ignore deprecation and future, and user
warnings.'''\nimport warnings as wrn\nwrn.filterwarnings('ignore',
category = DeprecationWarning) \nwrn.filterwarnings('ignore', category =
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town_state_df.Town.apply(lambda x: ' '.join(x.split('
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2002      \\n2      1112      2004 AG. CUAUTITLAN ESTADO DE MÉXICO      2004
\\n3      1113      2008 AG. LAGO FILT      MÉXICO, D.F.      2008      \\n4
1114      2029 AG. IZTAPALAPA 2      MÉXICO, D.F.      2029      \\n\\n
Town_name  \\n0      AG. LAGO FILT  \\n1      AG. AZCAPOTZALCO  \\n2      AG.
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Burritos Sincro 170g CU LON 53   \n4          72          Div Tira Mini
Doradita 4p 45g TR 72   \n\n          popular_name property unit
pieces   \n0          NaN          NaN          NaN   \n1
Capuccino Moka          NES    750g          NaN   \n2  Bimbollos Ext sAjonjoli
BIM    480g          6p   \n3          Burritos Sincro    CU LON    170g    NaN
\n4   Div Tira Mini Doradita          TR    45g    4p
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72</td>\n    <td>Div Tira Mini Doradita</td>\n    <td>TR</td>\n
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1\n2          2          SIN NOMBRE          1\n3
3          EL MORENO          1\n4          4 SDN
SER DE ALIM CUERPO SA CIA DE INT          2", "text/html": "<div>\n<style
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1226    4705135    1238\n2    2    10    2045    1
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[illegible]

[illegible]

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sale_price_sr\ndel return_price_sr\n\n\nprint(f'product price\\'s shape:
{product_price_df.shape}')\nproduct_price_df.head(5)", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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cross validation'''
\nkf = KFold(n_splits=3, shuffle=True,
random_state=2)\n\n\n'''Define the validation function'''
\ndef
rmsle_cv(model, X, y, cv=kf):\n    rmsle = np.sqrt(\n
cross_val_score(\n
model,\n
X, y,\n
scoring=\n\neg_mean_squared_log_error\n\n, \n
cv=cv,\n
)\n
)\n    return(rmsle)", "metadata": {"execution": {"iopub.status.busy": "2023-
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Cleansing", "metadata": {}}, {"cell_type": "markdown", "source": "###
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own'] = town_state_df['Town'].str.upper()\ntown_state_df['Town_name'] =
town_state_df['Town_name'].str.upper()\ntown_state_df['State'] =
town_state_df['State'].str.upper()", "metadata": {"execution": {"iopub.statu
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.isin(['LOS MOCHIS',
'PINOTEPA']))].sort_values(by='Town_name')", "metadata": {"execution": {"iop
ub.status.busy": "2023-03-28T02:52:45.465099Z", "iopub.status.idle": "2023-
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[2561, 2169,
2152])].sort_values(by='Town_ID')", "metadata": {"execution": {"iopub.status

```



```
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3217\ntown_state_df.loc[311, 'Town_ID'] =
3218","metadata":{"execution":{"iopub.status.busy":"2023-03-
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f = product_price_df.drop('return_price',
axis=1).dropna().rename(columns={'sale_price':
'price'})\nreturn_prices_df = product_price_df.drop('sale_price',
axis=1).dropna().rename(columns={'return_price': 'price'})\nprices_df =
pd.concat([sale_prices_df, return_prices_df])\nprices_df =
prices_df.groupby('Producto_ID')['price'].median().reset_index()\nprices_
df.head(5)","metadata":{"execution":{"iopub.status.busy":"2023-03-
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0\nproduct_df.loc[product_df['Producto_ID'].isin(test_df['Producto_ID']).u
nique()), 'in_test'] =
1\nproduct_df.loc[product_df['Producto_ID'].isin(train_df['Producto_ID']).
unique()), 'in_train'] = 1\nproduct_df =
product_df[(product_df['in_test'] == 1) | (product_df['in_train'] ==
1)]","metadata":{"execution":{"iopub.status.busy":"2023-03-
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'Donas'\nproduct_df.loc[117, 'property'] = 'Prom
BIM'\nproduct_df.loc[117, 'unit'] = None\nproduct_df.loc[117, 'pieces'] =
'6p'\n\nproduct_df.loc[190, 'popular_name'] = 'Paletina para
Cafe'\nproduct_df.loc[190, 'property'] = 'NES'\nproduct_df.loc[190,
'unit'] = None\nproduct_df.loc[190, 'pieces'] =
None\n\nproduct_df.loc[381, 'popular_name'] = 'Camioncitos
Bimbo'\nproduct_df.loc[381, 'property'] = 'BIM'\nproduct_df.loc[381,
'unit'] = None\nproduct_df.loc[381, 'pieces'] =
None\n\nproduct_df.loc[1152, 'popular_name'] = 'Burrito Vaporero
FrijolChorizo'\nproduct_df.loc[1152, 'property'] = 'CU
LON'\nproduct_df.loc[1152, 'unit'] = '90g'\nproduct_df.loc[1152,
'pieces'] = None\n\nproduct_df.loc[1677, 'popular_name'] = 'Tarima Twin
```

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Pack Thins Multig'\nproduct_df.loc[1677, 'property'] = 'CU
ORO'\nproduct_df.loc[1677, 'unit'] = None\nproduct_df.loc[1677, 'pieces']
= None\n\nproduct_df.loc[1888, 'popular_name'] = 'Deliciosas
Chochochispas'\nproduct_df.loc[1888, 'property'] = 'Prom MTA
LAR'\nproduct_df.loc[1888, 'unit'] = '204g'\nproduct_df.loc[1888,
'pieces'] = None\n\nproduct_df.loc[1889, 'popular_name'] = 'Deliciosas
Chochochispas'\nproduct_df.loc[1889, 'property'] = 'Prom
LAR'\nproduct_df.loc[1889, 'unit'] = '204g'\nproduct_df.loc[1889,
'pieces'] = None\n\nproduct_df.loc[2449, 'popular_name'] = 'Galleta
Granel Classics Chocolate'\nproduct_df.loc[2449, 'property'] =
'GBI'\nproduct_df.loc[2449, 'unit'] = None\nproduct_df.loc[2449,
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'535g'\nproduct_df.loc[877, 'pieces'] = '10p'\n\nproduct_df.loc[1585,
'popular_name'] = 'Principe Cho Bco MG'\nproduct_df.loc[1585, 'unit'] =
'110g'\nproduct_df.loc[1585, 'pieces'] = '10p'\n\nproduct_df.loc[1748,
'popular_name'] = 'Combo Salma mas Levite'\nproduct_df.loc[1748, 'unit']
= '1360g'\nproduct_df.loc[1748, 'pieces'] =
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product_df['pieces'].str.extract(r'(\d+) (p|Reb) ')[0]", "metadata": {"execu
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product_df['unit'].str.strip()\nproduct_df['weight'] =
product_df['weight'].str.replace(' ', '.')\nproduct_df['weight'] =
product_df['weight'].str.upper()\nw =
product_df['weight'].str.extract('(.*?) (KG|G|ML) ',
expand=True)\nproduct_df['weight'] = w[0].astype('float') *
w[1].map({'KG':1000, 'G':1,
'ML':1})", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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prices_df,
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] =
product_df['pieces'].fillna(1)", "metadata": {"execution": {"iopub.status.bu
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```

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28T02:52:45.497764Z"},"trusted":true},"execution_count":null,"outputs":[]
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}, {"cell_type": "code", "source": "df = product_df.dropna()\ndf =
df[(df['price'] <= 311) & (df['weight'] <=
1880)]\nplt.figure(figsize=(16,8))\nsns.scatterplot(x='weight',
y='price', data=df)\ndel
df", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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LinearRegression\n\ndf = product_df.dropna()\ndf = df[(df['price'] <=
100) & (df['weight'] <= 1880)]\n\n# predict missing prices\nlrf =
LinearRegression()\nlrf.fit(df['weight'].values.reshape(-1, 1),
df['price'])\n\nprices =
lrf.predict(product_df[product_df['price'].isnull()][ 'weight'].values.resh
ape(-1, 1))\n\nproduct_df.loc[product_df['price'].isnull(), 'price'] =
prices\n\n# predict missing weights\nlrf =
LinearRegression()\nlrf.fit(df['price'].values.reshape(-1, 1),
df['weight'])\n\nweights =
lrf.predict(product_df[product_df['weight'].isnull()][ 'price'].values.resh
ape(-1, 1))\n\nproduct_df.loc[product_df['weight'].isnull(), 'weight'] =
weights\n\ndel df\ndel prices\ndel
weights", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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axis=1).isnull().sum() ==
0).all()", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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Type", "metadata": {}}, {"cell_type": "code", "source": "product_df['pieces'] =
product_df['pieces'].astype('int16')\nproduct_df['in_train'] =
product_df['in_train'].astype('bool')\nproduct_df['in_test'] =
product_df['in_test'].astype('bool')\nproduct_df['weight'] =
product_df['weight'].astype('float32')\nproduct_df['price'] =
product_df['price'].astype('float32')", "metadata": {"execution": {"iopub.st
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train_df['Canal_ID'].astype('category')\ntest_df['Canal_ID'] =
test_df['Canal_ID'].astype('category')", "metadata": {"execution": {"iopub.s
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Features", "metadata": {}}, {"cell_type": "code", "source": "train_df.drop(['Ve
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inplace=True)", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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train_df.groupby('Ruta_SAK')['Demanda_uni_equil'].median().astype('int16'
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Data","metadata":{}},{ "cell_type":"code","source":"test_df =
pd.merge(test_df, client_med_s.reset_index(), how='left')\ntest_df =
pd.merge(test_df, popular_name_med_s.reset_index(), how='left')\ntest_df
= pd.merge(test_df, town_id_med_s.reset_index(), how='left')\ntest_df =
pd.merge(test_df, ruta_id_med_s.reset_index(),
how='left')","metadata":{"execution":{"iopub.status.busy":"2023-03-
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popular_name_med_s.reset_index(), how='left')\ntrain_df =
pd.merge(train_df, town_id_med_s.reset_index(), how='left')\ntrain_df =
pd.merge(train_df, ruta_id_med_s.reset_index(),
how='left')","metadata":{"execution":{"iopub.status.busy":"2023-03-
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Data","metadata":{}},{ "cell_type":"code","source":"test_df['popular_name_
med'] =
test_df['popular_name_med'].fillna(test_df['popular_name_med'].mean())\nt
est_df['client_med'] =
test_df['client_med'].fillna(test_df['client_med'].mean())\ntest_df['ruta
_id_med'] =
test_df['ruta_id_med'].fillna(test_df['ruta_id_med'].mean())","metadata":
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Type","metadata":{}},{ "cell_type":"code","source":"train_df['client_med']
= train_df['client_med'].astype('int16')\ntrain_df['popular_name_med'] =
train_df['popular_name_med'].astype('int16')\ntrain_df['town_id_med'] =
train_df['town_id_med'].astype('int16')\ntrain_df['ruta_id_med'] =
train_df['ruta_id_med'].astype('int16')","metadata":{"execution":{"iopub.s
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},{ "cell_type":"code","source":"test_df['client_med'] =
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test_df['popular_name_med'].astype('int16')\ntest_df['town_id_med'] =
test_df['town_id_med'].astype('int16')\ntest_df['ruta_id_med'] =
test_df['ruta_id_med'].astype('int16')","metadata":{"execution":{"iopub.s
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Features","metadata":{}},{ "cell_type":"code","source":"train_df.drop(\n
['Ruta_SAK', 'Cliente_ID', 'Producto_ID', 'Town_ID', 'popular_name',
'property', 'pieces'],\n    axis=1, inplace=True)\ntest_df.drop(\n

```

```

['Ruta_SAK', 'Cliente_ID', 'Producto_ID', 'Town_ID', 'popular_name',
'property', 'pieces'],\n      axis=1,
inplace=True)", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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Preprocessing", "metadata": {}}, {"cell_type": "markdown", "source": "## 3.1
Take a glance at all
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28T02:52:45.545016Z"}, "trusted": true}, "execution_count": null, "outputs": []
}, {"cell_type": "code", "source": "test_df.describe()", "metadata": {"executio
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28T02:52:45.546859Z"}, "trusted": true}, "execution_count": null, "outputs": []
}, {"cell_type": "code", "source": "'''Plot histogram of numerical variables
to validate pandas intuition.''\ndef draw_histograms(df, variables,
n_rows, n_cols, size):\n    fig=plt.figure()\n    for i, var_name in
enumerate(variables):\n        ax=fig.add_subplot(n_rows, n_cols, i+1)\n
df[var_name].hist(bins=40, ax=ax, color='skyblue', alpha=0.8,
figsize=size)\n        ax.set_title(var_name, fontsize=43)\n
ax.tick_params(axis='both', which='major', labelsize=35)\n
ax.tick_params(axis='both', which='minor', labelsize=35)\n
ax.set_xlabel('')\n    fig.tight_layout(rect=[0, 0.03, 1, 0.95])\n
plt.show()", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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}, {"cell_type": "code", "source": "skewness =
train_df.select_dtypes(include=['int8', 'int16', 'int32', 'int64',
'float32', 'float64']).apply(lambda x: skew(x))\nskew_index =
skewness[abs(skewness) >=
0.75].index\nnskewness[skew_index].sort_values(ascending=False)", "metadata
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test_df.select_dtypes(include=['int8', 'int16', 'int32', 'int64',
'float32', 'float64']).apply(lambda x: skew(x))\nskew_index =
skewness[abs(skewness) >=
0.75].index\nnskewness[skew_index].sort_values(ascending=False)", "metadata
": {"execution": {"iopub.status.busy": "2023-03-

```

```
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```

```

LightGBM'''\nimport lightgbm as lgb\nmodel_lgb =
lgb.LGBMRegressor(num_leaves=1000,\n
max_depth=5,\n
learning_rate=0.1,\n
random_state=2)\nmodel_lgb.fit(sample_train_df.drop(['Demanda_uni_equil',
'log_target'], axis=1),
sample_train_df['log_target']),"metadata":{"execution":{"iopub.status.bu
sy":"2023-03-28T02:52:45.569552Z","iopub.status.idle":"2023-03-
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pd.DataFrame(model_lgb.feature_importances_,\n
index=sample_train_df.drop(['Demanda_uni_equil', 'log_target'],
axis=1).columns,\n
columns=['importance']).sort_values('importance',
ascending=False)\nndf[df.importance >
10]","metadata":{"execution":{"iopub.status.busy":"2023-03-
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28T02:52:45.571813Z"},"trusted":true},"execution_count":null,"outputs":[]
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8))\nsns.scatterplot(x='weight', y='log_target', data=sample_train_df,
palette='Blues_d')","metadata":{"execution":{"iopub.status.busy":"2023-
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palette='Blues_d')","metadata":{"execution":{"iopub.status.busy":"2023-
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},{ "cell_type":"code","source":"#
popular_name_med\nplt.figure(figsize=(16,
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data=sample_train_df,
palette='Blues_d')","metadata":{"execution":{"iopub.status.busy":"2023-
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8))\nsns.scatterplot(x='town_id_med', y='log_target',
data=sample_train_df,
palette='Blues_d')","metadata":{"execution":{"iopub.status.busy":"2023-
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28T02:52:45.580903Z"},"trusted":true},"execution_count":null,"outputs":[]
},{ "cell_type":"code","source":"# ruta_id_med\nplt.figure(figsize=(16,
8))\nsns.scatterplot(x='ruta_id_med', y='log_target',
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Evaluation","metadata":{}},{ "cell_type":"markdown","source":"## 5.1
Importing
Packages","metadata":{}},{ "cell_type":"code","source":"'''Importing
Modeling Interested Modules'''\nfrom sklearn.base import
BaseEstimator\nfrom sklearn.pipeline import make_pipeline\nfrom

```



```

sklearn.preprocessing import StandardScaler, RobustScaler\nfrom
sklearn.model_selection import GridSearchCV\nfrom sklearn.linear_model
import LinearRegression, LassoCV, RidgeCV, ElasticNetCV\nfrom sklearn.svm
import SVR\nfrom sklearn.kernel_ridge import KernelRidge\nfrom lightgbm
import LGBMRegressor", "metadata": {"execution": {"iopub.status.busy": "2023-
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Models ", "metadata": {}}, {"cell_type": "code", "source": "'Adjust dataframe
for modeling'\n\ntrain_y =
train_df['Demanda_uni_equil']\ntrain_df.drop(['Demanda_uni_equil'],
axis=1, inplace=True)\ntrain_X = train_df\ntest_X = test_df.drop('id',
axis=1)\n\n'Transform categorical features to dummy
variables'\n\ntrain_X = pd.get_dummies(train_X)\ntest_X =
pd.get_dummies(test_X)", "metadata": {"execution": {"iopub.status.busy": "202
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training'\n\nsample_train_y =
sample_train_df['Demanda_uni_equil']\nsample_train_df.drop(['Demanda_uni_
equil', 'log_target'], axis=1, inplace=True)\nsample_train_X =
sample_train_df\n\nsample_train_X =
pd.get_dummies(sample_train_X)", "metadata": {"execution": {"iopub.status.bu
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}, {"cell_type": "code", "source": "'We should use the log transform of the
target value'\n\nclass MyEstimator(BaseEstimator):\n    def
__init__(self, model):\n        self.model = model\n\n    def
fit(self, X, y):\n        self.model.fit(X, np.log1p(y))\n\n    def
self\n\n    def predict(self, X):\n        predicts =
np.expml(self.model.predict(X))\n        mask = (predicts <= 0)\n\n    predicts[mask] = 0\n\n    return
predicts", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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28T02:52:45.594294Z"}, "trusted": true}, "execution_count": null, "outputs": []
}, {"cell_type": "code", "source": "'Define evaluation function for
Convenience'\n\ndef evaluation_model(model, train_X, train_y, test_X):\n
cv = rmsle_cv(model, train_X, train_y)\n    cv_mean = np.round(cv.mean(),
5)\n    cv_std = np.round(cv.std(), 5)\n    sample_prediction =
model.predict(test_X.loc[:, :])\n    return {'cv_mean': cv_mean,
'cv_std': cv_std, 'sample_prediction':
sample_prediction}", "metadata": {"execution": {"iopub.status.busy": "2023-
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28T02:52:45.596087Z"}, "trusted": true}, "execution_count": null, "outputs": []
}, {"cell_type": "code", "source": "'Define Hyperparameters Tuning
Function'\n\ndef tune_hyperparameters(model, param_grid, train_X,
train_y):\n    grid = GridSearchCV(\n        model, param_grid, \n
scoring='neg_mean_squared_log_error',\n        cv=3, n_jobs=-1,\n    )\n
grid.fit(train_X, train_y)\n    best_params = grid.best_params_ \n

```

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best_score = np.round(np.sqrt(-1 * grid.best_score_), 5)\n    return
best_params,
best_score", "metadata": {"execution": {"iopub.status.busy": "2023-03-
28T02:52:45.597310Z", "iopub.status.idle": "2023-03-
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Models", "metadata": {}}, {"cell_type": "markdown", "source": "###
LinearRegression", "metadata": {}}, {"cell_type": "code", "source": "model =
make_pipeline(\n    RobustScaler(),\n    LinearRegression(),\n)\n\nnlr_model =
MyEstimator(model)\n\nnlr_model.fit(sample_train_X,
sample_train_y)\n\nnlr_eval = evaluation_model(nlr_model, sample_train_X,
sample_train_y,
test_X)\n\nprint(nlr_eval)", "metadata": {"execution": {"iopub.status.busy": "20
23-03-28T02:52:45.599498Z", "iopub.status.idle": "2023-03-
28T02:52:45.600172Z"}, "trusted": true}, "execution_count": null, "outputs": []
}, {"cell_type": "markdown", "source": "###
LassoCV", "metadata": {}}, {"cell_type": "code", "source": "model =
make_pipeline(\n    RobustScaler(),\n    LassoCV(\n
alphas=(0.00001, 0.0001, 0.0005, 0.001, 0.01, 0.05, 0.1, 0.3, 1, 3, 5,
10),\n    ),\n)\n\nnl_lasso_cv_model =
MyEstimator(model)\n\nnl_lasso_cv_model.fit(sample_train_X,
sample_train_y)\n\nnl_lasso_cv_eval = evaluation_model(lasso_cv_model,
sample_train_X, sample_train_y,
test_X)\n\nprint(lasso_cv_eval)\n\n\nnopt_alpha =
lasso_cv_model.model.steps[1][1].alpha_\n\nprint(f'\nnopt_alpha:
{opt_alpha}')", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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RidgeCV", "metadata": {}}, {"cell_type": "code", "source": "model =
make_pipeline(\n    RobustScaler(),\n    RidgeCV(\n
alphas=(0.0001, 0.0005, 0.001, 0.01, 0.05, 0.1, 0.3, 1, 3, 5, 10),\n
),\n)\n\nnl_ridge_cv_model =
MyEstimator(model)\n\nnl_ridge_cv_model.fit(sample_train_X,
sample_train_y)\n\nnl_ridge_cv_eval = evaluation_model(ridge_cv_model,
sample_train_X, sample_train_y,
test_X)\n\nprint(ridge_cv_eval)\n\n\nnopt_alpha =
ridge_cv_model.model.steps[1][1].alpha_\n\nprint(f'\nnopt_alpha:
{opt_alpha}')", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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ElasticNetCV", "metadata": {}}, {"cell_type": "code", "source": "model =
make_pipeline(\n    RobustScaler(),\n    ElasticNetCV(\n
alphas=(0.00001, 0.0001, 0.0002, 0.0003), \n    l1_ratio=(0.001,
0.01, 0.1, 0.2, 0.3, 0.4, 0.5),\n    ),\n)\n\nnl_elastic_cv_model =
MyEstimator(model)\n\nnl_elastic_cv_model.fit(sample_train_X,
sample_train_y)\n\nnl_elastic_cv_eval = evaluation_model(elastic_cv_model,
sample_train_X, sample_train_y,
test_X)\n\nprint(elastic_cv_eval)\n\n\nnopt_alpha =
elastic_cv_model.model.steps[1][1].alpha_\n\nnopt_l1_ratio =
elastic_cv_model.model.steps[1][1].l1_ratio_\n\nprint(f'\nnopt_alpha:
{opt_alpha} nopt_l1_ratio:
{opt_l1_ratio}')", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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SVR", "metadata": {}}, {"cell_type": "code", "source": "# {'cv_mean': 0.57025,

```

```

'cv_std': 0.00775, 'sample_prediction': array([3.85650165, 1.24517075,
3.45866592])}\n\n# grid best_params: {'model__svr__C': 10,
'model__svr__epsilon': 0.1, 'model__svr__gamma': 0.01}\n\n# ### build
basemodel\n# model = make_pipeline(\n#     RobustScaler(),\n#
SVR(),\n# )\n# svr_model = MyEstimator(model)\n# \n# ### optimize
hyperparameters\n# param_grid = {'model__svr__C': [1, 10, 20],\n#
'model__svr__epsilon': [0.001, 0.01, 0.1],\n#
'model__svr__gamma': [0.0001, 0.001, 0.01]}\n# best_params, best_score =
\\n#     tune_hyperparameters(svr_model, param_grid, sample_train_X,
sample_train_y)\n# \n# ### fit using best_params\n#
svr_model.set_params(**best_params)\n# svr_model.fit(sample_train_X,
sample_train_y)\n# svr_eval = evaluation_model(svr_model, sample_train_X,
sample_train_y, test_X)\n# print(svr_eval)\n# \n# print(f'\ngrid
best_params:
{best_params}')", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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KernelRidge", "metadata": {}, {"cell_type": "code", "source": "# {'cv_mean':
0.56925, 'cv_std': 0.00546, 'sample_prediction': array([4.15061605,
1.21544493, 3.65678685, 1.81862787])}\n\n# grid best_params:
{'model__kernelridge__alpha': 0.5, 'model__kernelridge__coef0': 3,
'model__kernelridge__degree': 2, 'model__kernelridge__kernel':
'polynomial'}\n\n# ### build basemodel\n# model = make_pipeline(\n#
RobustScaler(),\n#     KernelRidge(),\n# )\n# kr_model =
MyEstimator(model)\n# \n# ### optimize hyperparameters\n# param_grid =
{'model__kernelridge__alpha': [0.01, 0.1, 0.5, 1],\n#
'model__kernelridge__kernel': ['linear', 'polynomial'],\n#
'model__kernelridge__degree': [1, 1.5, 2, 3],\n#
'model__kernelridge__coef0': [3, 4, 5]}\n# best_params, best_score =
\\n#     tune_hyperparameters(kr_model, param_grid, sample_train_X,
sample_train_y)\n# \n# ### fit using best_params\n#
kr_model.set_params(**best_params)\n# kr_model.fit(sample_train_X,
sample_train_y)\n# kr_eval = evaluation_model(kr_model, sample_train_X,
sample_train_y, test_X)\n# print(kr_eval)\n# \n# print(f'\ngrid
best_params:
{best_params}')", "metadata": {"execution": {"iopub.status.busy": "2023-03-
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LightGBM", "metadata": {}, {"cell_type": "code", "source": "model =
LGBMRegressor(learning_rate=0.01, n_estimators=3000,\n
num_leaves=5,\n
max_bin=55, bagging_fraction=0.8,\n
feature_fraction=0.2319,\n
bagging_seed=9,\n
min_data_in_leaf=6,\n
min_sum_hessian_in_leaf=11)\nlgb_model =
MyEstimator(model)\nlgb_model.fit(sample_train_X,
sample_train_y)\nlgb_eval = evaluation_model(lgb_model, sample_train_X,
sample_train_y,
test_X)\nprint(lgb_eval)", "metadata": {"execution": {"iopub.status.busy": "2
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Submission", "metadata": {}, {"cell_type": "code", "source": "def
output_submission_file(model, test_X, filename='submission.csv'):\n
prediction = model.predict(test_X)\n    df = pd.DataFrame({'id':
test_df['id'], 'Demanda_uni_equil': prediction})\n

```

```
print(f'{df.shape}')\n    print(f'{df.head(5)}')\n    df.to_csv(filename,\nindex=False)\n    df.to_csv(filename + '.gz', index=False,\ncompression='gzip'), "metadata":{"execution":{"iopub.status.busy":"2023-03-28T02:52:45.610677Z","iopub.status.idle":"2023-03-28T02:52:45.611166Z"},"trusted":true},"execution_count":null,"outputs":[]}, {"cell_type":"code","source":"'''\nSubmission'''\n\noutput_submission_file(\n    lgb_model, test_X\n)", "metadata":{"execution":{"iopub.status.busy":"2023-03-28T02:52:45.612678Z","iopub.status.idle":"2023-03-28T02:52:45.613290Z"},"trusted":true},"execution_count":null,"outputs":[]}\n    ]}]}
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