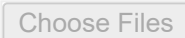


▼ ASHRAE - Energy Prediction III

```
from google.colab import files
```

```
uploaded = files.upload()
!mkdir -p ~/.kaggle/ && mv kaggle.json ~/.kaggle/ && chmod 600 ~/.kaggle/kaggle.json
!kaggle competitions download -c ashrae-energy-prediction
!unzip train.csv.zip
!unzip test.csv.zip
!unzip weather_test.csv.zip
!unzip weather_train.csv.zip
!unzip sample_submission.csv.zip
```

 No file chosen Upload widget is only available when the cell has been executed in

Saving kaggle.json to kaggle.json
Warning: Looks like you're using an outdated API Version, please consider updating (serv
Downloading train.csv.zip to /content
94% 113M/120M [00:01<00:00, 44.2MB/s]
100% 120M/120M [00:01<00:00, 71.1MB/s]
Downloading weather_test.csv.zip to /content
0% 0.00/2.53M [00:00<?, ?B/s]
100% 2.53M/2.53M [00:00<00:00, 83.8MB/s]
Downloading sample_submission.csv.zip to /content
100% 88.4M/88.4M [00:02<00:00, 37.3MB/s]

Downloading building_metadata.csv to /content
0% 0.00/44.5k [00:00<?, ?B/s]
100% 44.5k/44.5k [00:00<00:00, 44.2MB/s]
Downloading weather_train.csv.zip to /content
0% 0.00/1.27M [00:00<?, ?B/s]
100% 1.27M/1.27M [00:00<00:00, 179MB/s]
Downloading test.csv.zip to /content
91% 152M/167M [00:02<00:00, 34.7MB/s]
100% 167M/167M [00:03<00:00, 57.9MB/s]
Archive: train.csv.zip
inflating: train.csv
Archive: test.csv.zip
inflating: test.csv
Archive: weather_test.csv.zip
inflating: weather_test.csv
Archive: weather_train.csv.zip
inflating: weather_train.csv
Archive: sample_submission.csv.zip
inflating: sample_submission.csv

```
# upload external building data
building_data = files.upload()
```



No file chosen

Upload widget is only available when the cell has been executed in

```

import numpy as np
import pandas as pd
from sklearn.preprocessing import LabelEncoder, StandardScaler, OneHotEncoder
from xgboost import XGBRegressor
from sklearn.ensemble import GradientBoostingRegressor
from sklearn.model_selection import train_test_split, GridSearchCV
import gc, warnings, datetime
from sklearn.linear_model import LinearRegression, Ridge, Lasso, ElasticNet
from sklearn.metrics import mean_squared_error
from math import sqrt
import seaborn as sns
import matplotlib.pyplot as plt
# Create lightgbm model
import lightgbm as lgb
from sklearn.model_selection import KFold, StratifiedKFold
from tqdm import tqdm
warnings.filterwarnings("ignore")

# Load dataframes
train = pd.read_csv('train.csv')
test = pd.read_csv('test.csv')
# building = pd.read_csv('building_metadata.csv')
building = pd.read_csv('building_metadata_external.csv') # using external data with 2 new col
weather_train = pd.read_csv('weather_train.csv')
weather_test = pd.read_csv('weather_test.csv')

# Separate row_ids for submission file
row_ids = test['row_id']
building.drop(['leed'], inplace=True, axis=1)
test.drop('row_id', inplace=True, axis=1)

'''Function to reduce the DF size'''
# source: https://www.kaggle.com/kernels/scriptcontent/3684066/download

def reduce_mem_usage(df):
    """ iterate through all the columns of a dataframe and modify the data type
    to reduce memory usage.
    """
    start_mem = df.memory_usage().sum() / 1024**2
    print('Memory usage of dataframe is {:.2f} MB'.format(start_mem))

    for col in df.columns:
        col_type = df[col].dtype

        if col_type != object:
            c_min = df[col].min()
            c_max = df[col].max()
            if str(col_type)[:3] == 'int':

```

```

    if c_min > np.iinfo(np.int8).min and c_max < np.iinfo(np.int8).max:
        df[col] = df[col].astype(np.int8)
    elif c_min > np.iinfo(np.int16).min and c_max < np.iinfo(np.int16).max:
        df[col] = df[col].astype(np.int16)
    elif c_min > np.iinfo(np.int32).min and c_max < np.iinfo(np.int32).max:
        df[col] = df[col].astype(np.int32)
    elif c_min > np.iinfo(np.int64).min and c_max < np.iinfo(np.int64).max:
        df[col] = df[col].astype(np.int64)
    else:
        if c_min > np.finfo(np.float16).min and c_max < np.finfo(np.float16).max:
            df[col] = df[col].astype(np.float16)
        elif c_min > np.finfo(np.float32).min and c_max < np.finfo(np.float32).max:
            df[col] = df[col].astype(np.float32)
        else:
            df[col] = df[col].astype(np.float64)
    else:
        if col == 'timestamp':
            print('Keeping timestamp...')
        else:
            df[col] = df[col].astype('category')

end_mem = df.memory_usage().sum() / 1024**2
print('Memory usage after optimization is: {:.2f} MB'.format(end_mem))
print('Decreased by {:.1f}%'.format(100 * (start_mem - end_mem) / start_mem))

return df

```

```

train = reduce_mem_usage(train)
test = reduce_mem_usage(test)

```

```

☞ Memory usage of dataframe is 616.95 MB
Keeping timestamp...
Memory usage after optimization is: 289.19 MB
Decreased by 53.1%
Memory usage of dataframe is 954.38 MB
Keeping timestamp...
Memory usage after optimization is: 437.43 MB
Decreased by 54.2%

```

```

# create column with log of readings for distribution normalization
train['meter_reading_log1p'] = np.log1p(train['meter_reading'])

```

```

# converting timestamp to datetime
train['timestamp'] = pd.to_datetime(train['timestamp'])
test['timestamp'] = pd.to_datetime(test['timestamp'])
weather_train['timestamp'] = pd.to_datetime(weather_train['timestamp'])
weather_test['timestamp'] = pd.to_datetime(weather_test['timestamp'])

```

```

# create date columns using train dataset
train['timestamp_hour'] = train.timestamp.dt.hour
train['timestamp_day'] = train.timestamp.dt.day # -> possibly overfitting
# train['timestamp_month'] = train.timestamp.dt.month -> possibly overfitting

```

```

# train['timestamp_year'] = train.timestamp.dt.year -> possibly overfitting
train['timestamp_weekday'] = train.timestamp.dt.weekday

# create date columns using test dataset
test['timestamp_hour'] = test.timestamp.dt.hour
test['timestamp_day'] = test.timestamp.dt.day # -> possibly overfitting
# test['timestamp_month'] = test.timestamp.dt.month -> possibly overfitting
# test['timestamp_year'] = test.timestamp.dt.year -> possibly overfitting
test['timestamp_weekday'] = test.timestamp.dt.weekday

# fill missing values in wheater train data set using interpolation
weather_train = weather_train.groupby('site_id').apply(lambda group: group.interpolate(limit_
weather_test = weather_test.groupby('site_id').apply(lambda group: group.interpolate(limit_di

# fill missing values in building_data set using interpolation
building = building.groupby('building_id').apply(lambda group: group.interpolate(limit_direct

# fill the remaining values with the median
b_cols = ['floor_count', 'year_built', 'eui']
for c in b_cols:
    building[c].fillna(building[c].median(), inplace=True)
b_cols = ['cloud_coverage', 'precip_depth_1_hr', 'sea_level_pressure']
for c in b_cols:
    weather_train[c].fillna(weather_train[c].median(), inplace=True)
    weather_test[c].fillna(weather_test[c].median(), inplace=True)

# merge building data
train = train.merge(building, on = 'building_id', how = 'left')
test = test.merge(building, on = 'building_id', how = 'left')
# merge weather data
train = train.merge(weather_train, on = ['site_id', 'timestamp'], how = 'left')
test = test.merge(weather_test, on = ['site_id', 'timestamp'], how = 'left')

# delete unnecessary data
del weather_test, weather_train, building
gc.collect()

0

# all eletricity values until March 20 for site_id == 0 are 0,so we'll remove them
to_remove = train.loc[(train.building_id <= 104) & (train.meter == 0) & (train.timestamp <= d
train.drop(to_remove.index,axis=0,inplace=True)

# we can drop the timestamp column now
train.drop(['timestamp'],axis=1,inplace=True)
test.drop(['timestamp'],axis=1,inplace=True)

# encode primary_use column using LabelEncoder
le = LabelEncoder()

```

```

train.loc[:, 'primary_use'] = le.fit_transform(train.primary_use)
test.loc[:, 'primary_use'] = le.fit_transform(test.primary_use)

#We can drop null rows 'cause they mean that there is no corresponding in the weather data fo
train.dropna(axis =0,inplace=True)

b_cols = ['cloud_coverage', 'precip_depth_1_hr', 'sea_level_pressure', 'dew_temperature', 'wind_d
for c in b_cols:
    test[c].fillna(test[c].median(), inplace=True)

# create X and y for model input
train = reduce_mem_usage(train)
y = train.meter_reading_log1p
X = train.drop(['meter_reading_log1p'], axis=1)
X.drop(['meter_reading'], axis=1, inplace=True)
# X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.33,random_state=42)

gc.collect()

☞ Memory usage of dataframe is 2923.78 MB
Memory usage after optimization is: 886.57 MB
Decreased by 69.7%
0

X.drop(['wind_direction', 'wind_speed', 'sea_level_pressure'], axis=1, inplace=True)

X.drop(['timestamp_month'], axis=1, inplace=True)

params = {
    'boosting_type': 'gbdt',
    'objective': 'regression',
    'metric': {'rmse'},
    # 'subsample': 0.25,
    # 'subsample_freq': 1,
    'learning_rate': 0.05, # best = .3
    'num_leaves': 40, # best = 4
    'feature_fraction': 0.7, # best = .7
    'reg_lambda': 2, # best = 1
    # 'lambda_l2': 1, # best = 1
    'max_depth': 8, #best = 2
    # 'bagging_fraction': 0.7,
    # 'min_data': 100
}

cat_feat = ["site_id", "building_id", "primary_use", "meter"]
folds = 4
seed = 55
kf = StratifiedKFold(n_splits=folds, shuffle=True, random_state=seed) # shuffle or not

def train_model(x .v ):

```

```
...
# oof_pred = np.zeros(train.shape[0]) # out of fold predictions
models = []
## stratify data by building_id
for train_index, val_index in tqdm(kf.split(x_, x_['meter'])):
    train_X = x_.iloc[train_index]
    val_X = x_.iloc[val_index]
    train_y = y_.iloc[train_index]
    val_y = y_.iloc[val_index]
    lgb_train = lgb.Dataset(train_X, train_y, categorical_feature=cat_feat)
    lgb_eval = lgb.Dataset(val_X, val_y, categorical_feature=cat_feat)
    gbm = lgb.train(params,
                    lgb_train,
                    num_boost_round=500,
                    valid_sets=(lgb_train, lgb_eval),
                    early_stopping_rounds=100,
                    verbose_eval = 100
                    )
    models.append(gbm)
    del lgb_train, lgb_eval
    gc.collect()
return models

# submission training
models = train_model(X,y)
```



```

0it [00:00, ?it/s]Training until validation scores don't improve for 100 rounds.
[100]   training's rmse: 1.18202          valid_1's rmse: 1.18181
[200]   training's rmse: 1.06138          valid_1's rmse: 1.06117
[300]   training's rmse: 1.0161 valid_1's rmse: 1.01594
[400]   training's rmse: 0.989771         valid_1's rmse: 0.989689

```

```

1it [06:07, 367.39s/it][500]   training's rmse: 0.974087          valid_1's rmse: 0.97407
Did not meet early stopping. Best iteration is:
[500]   training's rmse: 0.974087          valid_1's rmse: 0.97407
Training until validation scores don't improve for 100 rounds.
[100]   training's rmse: 1.18158          valid_1's rmse: 1.18231
[200]   training's rmse: 1.06528          valid_1's rmse: 1.06608
[300]   training's rmse: 1.02282          valid_1's rmse: 1.02375
[400]   training's rmse: 0.994377          valid_1's rmse: 0.995603

```

```

2it [12:24, 370.38s/it][500]   training's rmse: 0.977232          valid_1's rmse: 0.978638
Did not meet early stopping. Best iteration is:

```

```

df_fimp_1 = pd.DataFrame()
df_fimp_1["feature"] = X.columns.values
df_fimp_1["importance"] = models[0].feature_importance()
df_fimp_1["half"] = 1

```

```

df_fimp_2 = pd.DataFrame()
df_fimp_2["feature"] = X.columns.values
df_fimp_2["importance"] = models[1].feature_importance()
df_fimp_2["half"] = 2

```

```

df_fimp_3 = pd.DataFrame()
df_fimp_3["feature"] = X.columns.values
df_fimp_3["importance"] = models[2].feature_importance()
df_fimp_3["half"] = 2

```

```

df_fimp_4 = pd.DataFrame()
df_fimp_4["feature"] = X.columns.values
df_fimp_4["importance"] = models[3].feature_importance()
df_fimp_4["half"] = 2

```

```

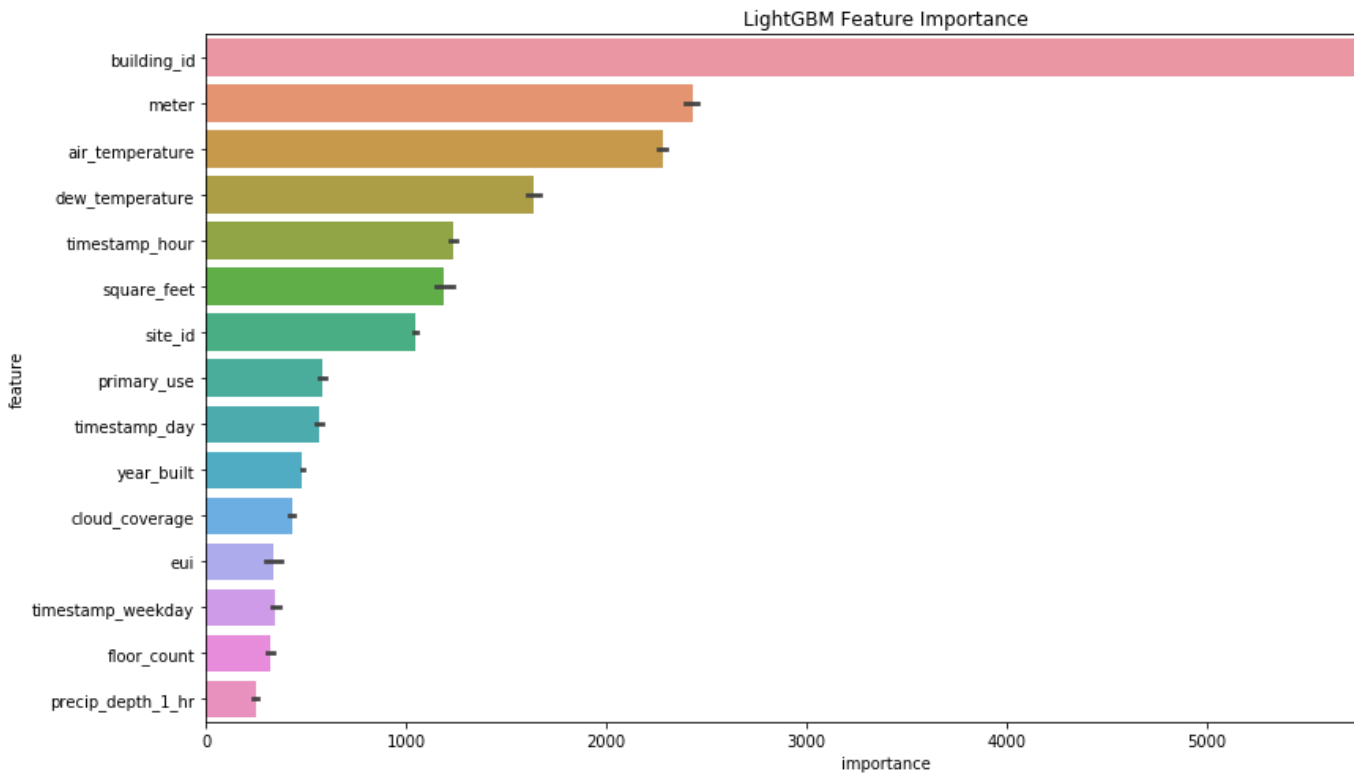
df_fimp = pd.concat([df_fimp_1, df_fimp_2, df_fimp_3, df_fimp_4], axis=0)

```

```

plt.figure(figsize=(14, 7))
sns.barplot(x="importance", y="feature", data=df_fimp.sort_values(by="importance", ascending=
plt.title("LightGBM Feature Importance")
plt.tight_layout()

```



```
#Predict using models
def predict_using_models(df):
    i=0
    result=[]
    step_size = 50000
    for j in tqdm(range(int(np.ceil(df.shape[0]/50000)))):
        result.append(np.expm1(sum([model.predict(df.iloc[i:i+step_size]) for model in models])))
        i+=step_size
    result = np.concatenate(result)
    return result

result = predict_using_models(test)
```




```
0%|          | 0/834 [00:00<?, ?it/s]

0%|          | 1/834 [00:03<53:48, 3.88s/it]

0%|          | 2/834 [00:07<54:00, 3.89s/it]

0%|          | 3/834 [00:11<54:02, 3.90s/it]

0%|          | 4/834 [00:15<53:58, 3.90s/it]

1%|          | 5/834 [00:19<53:54, 3.90s/it]

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1%|          | 9/834 [00:35<53:41, 3.91s/it]

1%|          | 10/834 [00:39<53:31, 3.90s/it]

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1%||         | 12/834 [00:46<53:20, 3.89s/it]

2%||         | 13/834 [00:50<53:24, 3.90s/it]
```

2%	14/834 [00:54<53:27, 3.91s/it]
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2%	18/834 [01:10<53:11, 3.91s/it]
2%	19/834 [01:14<53:04, 3.91s/it]
2%	20/834 [01:18<53:07, 3.92s/it]
3%	21/834 [01:22<52:58, 3.91s/it]
3%	22/834 [01:25<52:51, 3.91s/it]
3%	23/834 [01:29<52:50, 3.91s/it]
3%	24/834 [01:33<52:44, 3.91s/it]
3%	25/834 [01:37<52:49, 3.92s/it]
3%	26/834 [01:41<52:45, 3.92s/it]
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3% | 28/834 [01:49<52:41, 3.92s/it]

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4% | 31/834 [02:01<52:30, 3.92s/it]

4% | 32/834 [02:05<52:21, 3.92s/it]

4% | 33/834 [02:09<52:16, 3.92s/it]

4% | 34/834 [02:13<52:11, 3.91s/it]

4% | 35/834 [02:16<52:07, 3.91s/it]

4% | 36/834 [02:20<52:02, 3.91s/it]

4% | 37/834 [02:24<52:00, 3.92s/it]

5% | 38/834 [02:28<51:51, 3.91s/it]

5% | 39/834 [02:32<51:55, 3.92s/it]

5% | 40/834 [02:36<51:43, 3.91s/it]

5% | 41/834 [02:40<51:25, 3.89s/it]

5% | 42/834 [02:44<51:11, 3.88s/it]

5% | 42/834 [02:44<51:11, 3.88s/it]

5% | 43/834 [02:48<51:07, 3.88s/it]

5% | 44/834 [02:51<50:50, 3.86s/it]

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7% | ■ | 62/834 [04:01<49:34, 3.85s/it]

8% | ■ | 63/834 [04:04<49:24, 3.85s/it]

8% | ■ | 64/834 [04:08<49:25, 3.85s/it]

8% | ■ | 65/834 [04:12<49:22, 3.85s/it]

8% | ■ | 66/834 [04:16<49:10, 3.84s/it]

8% | ■ | 67/834 [04:20<49:15, 3.85s/it]

8% | ■ | 68/834 [04:24<49:09, 3.85s/it]

8% | ■ | 69/834 [04:28<49:09, 3.85s/it]

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9% | ■ | 74/834 [04:47<49:07, 3.88s/it]

9% | ■ | 75/834 [04:51<49:05, 3.88s/it]

9% | ■ | 76/834 [04:55<49:02, 3.88s/it]

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9% | ■ | 78/834 [05:03<48:49, 3.88s/it]

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10% | ■ | 82/834 [05:18<48:37, 3.88s/it]

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12% |  | 103/834 [06:40<47:26, 3.89s/it]

12% |  | 104/834 [06:44<47:30, 3.90s/it]

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13% |  | 106/834 [06:51<47:18, 3.90s/it]

13% |  | 107/834 [06:55<47:02, 3.88s/it]

13% |  | 108/834 [06:59<47:00, 3.89s/it]

13% |  | 109/834 [07:03<46:52, 3.88s/it]

13% |  | 110/834 [07:07<46:54, 3.89s/it]

13% |  | 111/834 [07:11<46:51, 3.89s/it]

13% |  | 112/834 [07:15<47:02, 3.91s/it]

14% |  | 113/834 [07:19<46:51, 3.90s/it]

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14% |  | 117/834 [07:34<46:31, 3.89s/it]

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14% |  | 119/834 [07:42<46:26, 3.90s/it]

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15% |  | 121/834 [07:50<46:12, 3.89s/it]

15% |  | 122/834 [07:54<46:13, 3.90s/it]

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15% |  | 125/834 [08:05<45:46, 3.87s/it]

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52% |  | 430/834 [27:16<24:52, 3.69s/it]

52% |  | 431/834 [27:19<24:47, 3.69s/it]

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84% |  | 697/834 [43:51<08:28, 3.71s/it]

84% |  | 698/834 [43:54<08:27, 3.73s/it]

84% |  | 699/834 [43:58<08:24, 3.74s/it]

84% |  | 700/834 [44:02<08:22, 3.75s/it]

84% |  | 701/834 [44:06<08:19, 3.76s/it]

84% |  | 702/834 [44:10<08:19, 3.78s/it]

84% |  | 703/834 [44:13<08:13, 3.77s/it]

84% |  | 704/834 [44:17<08:09, 3.77s/it]

85% | ██████████ | 705/834 [44:21<08:06, 3.77s/it]

85% | ██████████ | 706/834 [44:25<08:02, 3.77s/it]

85% | ██████████ | 707/834 [44:28<07:58, 3.77s/it]

85% | ██████████ | 708/834 [44:32<07:53, 3.76s/it]

85% | ██████████ | 709/834 [44:36<07:50, 3.76s/it]

85% | ██████████ | 710/834 [44:40<07:46, 3.76s/it]

85% | ██████████ | 711/834 [44:43<07:41, 3.75s/it]

85% | ██████████ | 712/834 [44:47<07:37, 3.75s/it]

85% | ██████████ | 713/834 [44:51<07:33, 3.75s/it]

86% | ██████████ | 714/834 [44:55<07:29, 3.75s/it]

86% | ██████████ | 715/834 [44:58<07:26, 3.75s/it]

86% | ██████████ | 716/834 [45:02<07:23, 3.76s/it]

86% | ██████████ | 717/834 [45:06<07:19, 3.76s/it]

86% | ██████████ | 718/834 [45:10<07:16, 3.76s/it]

86% | ██████████ | 719/834 [45:13<07:14, 3.78s/it]

86% | ██████████ | 720/834 [45:17<07:08, 3.76s/it]

86% | ██████████ | 721/834 [45:21<07:06, 3.77s/it]

87% | ██████████ | 722/834 [45:25<07:01, 3.76s/it]

87% | ██████████ | 723/834 [45:28<06:56, 3.75s/it]

87% | ██████████ | 724/834 [45:32<06:52, 3.75s/it]

87% | ██████████ | 725/834 [45:36<06:48, 3.75s/it]

87% | ██████████ | 726/834 [45:40<06:46, 3.76s/it]

87% | ██████████ | 727/834 [45:44<06:41, 3.76s/it]

87% | ██████████ | 728/834 [45:47<06:37, 3.75s/it]

87% | ██████████ | 729/834 [45:51<06:34, 3.76s/it]

88% | ██████████ | 730/834 [45:55<06:31, 3.77s/it]

88% | ██████████ | 731/834 [45:59<06:29, 3.78s/it]

88% | ██████████ | 732/834 [46:02<06:25, 3.78s/it]

88% | ██████████ | 733/834 [46:06<06:21, 3.77s/it]

88% |  | 734/834 [46:10<06:15, 3.76s/it]

88% |  | 735/834 [46:14<06:12, 3.77s/it]

88% |  | 736/834 [46:17<06:09, 3.77s/it]

88% |  | 737/834 [46:21<06:06, 3.77s/it]

88% |  | 738/834 [46:25<06:01, 3.77s/it]

89% |  | 739/834 [46:29<05:57, 3.76s/it]

89% |  | 740/834 [46:33<05:55, 3.78s/it]

89% |  | 741/834 [46:36<05:51, 3.78s/it]

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89% |  | 744/834 [46:48<05:40, 3.78s/it]

89% |  | 745/834 [46:51<05:37, 3.80s/it]

89% |  | 746/834 [46:55<05:34, 3.80s/it]

90% |  | 747/834 [46:59<05:32, 3.83s/it]

90% |  | 748/834 [47:03<05:30, 3.84s/it]

90% | ██████████ | 749/834 [47:07<05:27, 3.85s/it]

90% | ██████████ | 750/834 [47:11<05:22, 3.84s/it]

90% | ██████████ | 751/834 [47:15<05:19, 3.85s/it]

90% | ██████████ | 752/834 [47:18<05:15, 3.85s/it]

90% | ██████████ | 753/834 [47:22<05:11, 3.84s/it]

90% | ██████████ | 754/834 [47:26<05:08, 3.85s/it]

91% | ██████████ | 755/834 [47:30<05:05, 3.87s/it]

91% | ██████████ | 756/834 [47:34<05:00, 3.86s/it]

91% | ██████████ | 757/834 [47:38<04:56, 3.86s/it]

91% | ██████████ | 758/834 [47:42<04:53, 3.86s/it]

91% | ██████████ | 759/834 [47:45<04:49, 3.86s/it]

91% | ██████████ | 760/834 [47:49<04:45, 3.86s/it]

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91% | ██████████ | 762/834 [47:57<04:38, 3.87s/it]

91% | ██████████ | 763/834 [48:01<04:34, 3.87s/it]

92% | ██████████ | 764/834 [48:05<04:31, 3.88s/it]

92% | ██████████ | 765/834 [48:09<04:27, 3.87s/it]

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92% | ██████████ | 767/834 [48:16<04:19, 3.88s/it]

92% | ██████████ | 768/834 [48:20<04:15, 3.86s/it]

92% | ██████████ | 769/834 [48:24<04:11, 3.86s/it]

92% | ██████████ | 770/834 [48:28<04:07, 3.86s/it]

92% | ██████████ | 771/834 [48:32<04:03, 3.87s/it]

93% | ██████████ | 772/834 [48:36<04:01, 3.89s/it]

93% | ██████████ | 773/834 [48:40<03:57, 3.89s/it]

93% | ██████████ | 774/834 [48:44<03:53, 3.89s/it]

93% | ██████████ | 775/834 [48:48<03:48, 3.88s/it]

93% | ██████████ | 776/834 [48:51<03:44, 3.87s/it]

93% | ██████████ | 777/834 [48:55<03:41, 3.88s/it]

93% | ██████████ | 777/834 [48:55<03:41, 3.86s/it]

93% | ██████████ | 778/834 [48:59<03:36, 3.87s/it]

93% | ██████████ | 779/834 [49:03<03:32, 3.87s/it]

94% | ██████████ | 780/834 [49:07<03:28, 3.87s/it]

94% | ██████████ | 781/834 [49:11<03:25, 3.87s/it]

94% | ██████████ | 782/834 [49:15<03:21, 3.87s/it]

94% | ██████████ | 783/834 [49:18<03:17, 3.87s/it]

94% | ██████████ | 784/834 [49:22<03:13, 3.86s/it]

94% | ██████████ | 785/834 [49:26<03:09, 3.87s/it]

94% | ██████████ | 786/834 [49:30<03:05, 3.86s/it]

94% | ██████████ | 787/834 [49:34<03:01, 3.86s/it]

94% | ██████████ | 788/834 [49:38<02:57, 3.85s/it]

95% | ██████████ | 789/834 [49:42<02:53, 3.85s/it]

95% | ██████████ | 790/834 [49:45<02:49, 3.86s/it]

95% | ██████████ | 791/834 [49:49<02:46, 3.87s/it]

95% |  | 792/834 [49:53<02:42, 3.86s/it]

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95% |  | 794/834 [50:01<02:34, 3.87s/it]

95% |  | 795/834 [50:05<02:31, 3.88s/it]

95% |  | 796/834 [50:09<02:27, 3.88s/it]

96% |  | 797/834 [50:13<02:23, 3.88s/it]

96% |  | 798/834 [50:17<02:19, 3.88s/it]

96% |  | 799/834 [50:20<02:15, 3.86s/it]

96% |  | 800/834 [50:24<02:11, 3.87s/it]

96% |  | 801/834 [50:28<02:07, 3.87s/it]

96% |  | 802/834 [50:32<02:03, 3.86s/it]

96% |  | 803/834 [50:36<01:59, 3.87s/it]

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98% |  | 819/834 [51:38<00:58, 3.88s/it]

98% |  | 820/834 [51:42<00:54, 3.88s/it]

98%|██████████ | 821/834 [51:46<00:50, 3.88s/it]

99%|██████████ | 822/834 [51:50<00:46, 3.88s/it]

```
# Concatenate predictions and create submission file
submission = pd.read_csv("sample_submission.csv")
submission["meter_reading"] = result
submission.to_csv("submission.csv", index = False)
```

```
!kaggle competitions submit -c ashrae-energy-prediction -f 'submission.csv' -m "LGBM"
```

```
⏏ Warning: Looks like you're using an outdated API Version, please consider updating (serv
100% 1.06G/1.06G [00:32<00:00, 34.9MB/s]
Successfully submitted to ASHRAE - Great Energy Predictor III
```