This notebook is an exercise in the <u>Intermediate Machine Learning</u> course. You can reference the tutorial at this link.

As a warm-up, you'll review some machine learning fundamentals and submit your initial results to a Kaggle competition.

## Setup

The questions below will give you feedback on your work. Run the following cell to set up the feedback system.

```
In [ ]: # Set up code checking
        import os
        if not os.path.exists("../input/train.csv"):
            os.symlink("../input/home-data-for-ml-course/train.csv", "../input/t
            os.symlink("../input/home-data-for-ml-course/test.csv", "../input/te
        st.csv")
        from learntools.core import binder
        binder.bind(globals())
        from learntools.ml_intermediate.ex1 import *
        print("Setup Complete")
```

You will work with data from the Housing Prices Competition for Kaggle Learn Users to predict home prices in Iowa using 79 explanatory variables describing (almost) every aspect of the homes.

Ames Housing dataset image

Run the next code cell without changes to load the training and validation features in X\_train and X\_valid, along with the prediction targets in y\_train and y\_valid. The test features are loaded in X\_test . (If you need to review features and prediction targets, please check out this short tutorial. To read about model validation, look here. Alternatively, if you'd prefer to look through a full course to review all of these topics, start <u>here</u>.)

```
In [ ]: import pandas as pd
        from sklearn.model_selection import train_test_split
        # Read the data
        X_full = pd.read_csv('../input/train.csv', index_col='Id')
        X_test_full = pd.read_csv('../input/test.csv', index_col='Id')
        # Obtain target and predictors
        y = X_{full.SalePrice}
        features = ['LotArea', 'YearBuilt', '1stFlrSF', '2ndFlrSF', 'FullBath',
        'BedroomAbvGr', 'TotRmsAbvGrd']
        X = X_{full[features].copy()}
        X_test = X_test_full[features].copy()
        # Break off validation set from training data
        X_train, X_valid, y_train, y_valid = train_test_split(X, y, train_size=
        0.8, test_size=0.2,
                                                               random_state=0)
```

Use the next cell to print the first several rows of the data. It's a nice way to get an overview of the data you will use in your price prediction model.

```
In [ ]: X_train.head()
```

The next code cell defines five different random forest models. Run this code cell without changes. (To review **random forests**, look <u>here</u>.)

```
In [ ]: from sklearn.ensemble import RandomForestRegressor
        # Define the models
        model_1 = RandomForestRegressor(n_estimators=50, random_state=0)
        model_2 = RandomForestRegressor(n_estimators=100, random_state=0)
        model_3 = RandomForestRegressor(n_estimators=100, criterion='mae', rando
        m_state=0)
        model 4 = RandomForestRegressor(n_estimators=200, min_samples_split=20,
        random_state=0)
        model_5 = RandomForestRegressor(n_estimators=100, max_depth=7, random_st
        ate=0)
        models = [model_1, model_2, model_3, model_4, model_5]
```

To select the best model out of the five, we define a function score\_model() below. This function returns the mean absolute error (MAE) from the validation set. Recall that the best model will obtain the lowest MAE. (*To review mean absolute error*, look <u>here</u>.)

Run the code cell without changes.

```
In [ ]: from sklearn.metrics import mean_absolute_error
        # Function for comparing different models
        def score_model(model, X_t=X_train, X_v=X_valid, y_t=y_train, y_v=y_vali
            model.fit(X_t, y_t)
            preds = model.predict(X_v)
            return mean_absolute_error(y_v, preds)
        for i in range(0, len(models)):
            mae = score_model(models[i])
            print("Model %d MAE: %d" % (i+1, mae))
```

## **Step 1: Evaluate several models**

Use the above results to fill in the line below. Which model is the best model? Your answer should be one of model\_1, model\_2, model\_3, model\_4, or model\_5.

```
In [ ]: from sklearn.ensemble import RandomForestRegressor
        # Define the models
        model_1 = RandomForestRegressor(n_estimators=50, random_state=0)
        model_2 = RandomForestRegressor(n_estimators=100, random_state=0)
        model_3 = RandomForestRegressor(n_estimators=100, criterion='mae', rando
        m_state=0)
        model_4 = RandomForestRegressor(n_estimators=200, min_samples_split=20,
        random_state=0)
        model_5 = RandomForestRegressor(n_estimators=100, max_depth=7, random_st
        ate=0)
        models = [model_1, model_2, model_3, model_4, model_5]
In [ ]: from sklearn.metrics import mean_absolute_error
        # Function for comparing different models
        def score_model(model, X_t=X_train, X_v=X_valid, y_t=y_train, y_v=y_vali
        d):
            model.fit(X_t, y_t)
            preds = model.predict(X_v)
            return mean_absolute_error(y_v, preds)
        for i in range(0, len(models)):
            mae = score_model(models[i])
            print("Model %d MAE: %d" % (i+1, mae))
In [ ]: # Fill in the best model
        best_model = model_3
        # Check your answer
        step_1.check()
In [ ]: # Lines below will give you a hint or solution code
        #step_1.hint()
        #step_1.solution()
```

## **Step 2: Generate test predictions** Great. You know how to evaluate what makes an accurate model. Now it's time to go through the

modeling process and make predictions. In the line below, create a Random Forest model with the variable name my\_model.

```
my_model = RandomForestRegressor(random_state=0) # Your code here
         # Check your answer
         step_2.check()
In [ ]: # Lines below will give you a hint or solution code
         #step_2.hint()
         #step_2.solution()
         Run the next code cell without changes. The code fits the model to the training and validation
```

data, and then generates test predictions that are saved to a CSV file. These test predictions can be submitted directly to the competition! In [ ]: # Fit the model to the training data

```
# Generate test predictions
preds_test = my_model.predict(X_test)
# Save predictions in format used for competition scoring
output = pd.DataFrame({'Id': X_test.index,
                       'SalePrice': preds_test})
output.to_csv('submission.csv', index=False)
```

**Submit your results** 

Once you have successfully completed Step 2, you're ready to submit your results to the leaderboard! First, you'll need to join the competition if you haven't already. So open a new window by clicking on this link. Then click on the Join Competition button.

my\_model.fit(X, y)

In [ ]: |# Define a model

join competition image

- Next, follow the instructions below: 1. Begin by clicking on the blue **Save Version** button in the top right corner of the window. This
- will generate a pop-up window. 2. Ensure that the **Save and Run All** option is selected, and then click on the blue **Save** button.
- 3. This generates a window in the bottom left corner of the notebook. After it has finished running, click on the number to the right of the Save Version button. This pulls up a list of
- versions on the right of the screen. Click on the ellipsis (...) to the right of the most recent version, and select Open in Viewer. This brings you into view mode of the same page. You

4. Click on the **Output** tab on the right of the screen. Then, click on the file you would like to submit, and click on the blue Submit button to submit your results to the leaderboard.

You have now successfully submitted to the competition!

will need to scroll down to get back to these instructions.

If you want to keep working to improve your performance, select the blue **Edit** button in the top right of the screen. Then you can change your code and repeat the process. There's a lot of room to improve, and you will climb up the leaderboard as you work.

## **Keep going**

You've made your first model. But how can you quickly make it better?

Learn how to improve your competition results by incorporating columns with <u>missing values</u>.

Have questions or comments? Visit the <u>Learn Discussion forum</u> to chat with other Learners.