

After training our linear regression model, the next step is to utilize it for making predictions on new, unseen data. Follow these steps to guide you through the process:

Make predictions on new data

Make predictions on the new, normalized test data:

```
✓ [29] reg_pred = regression.predict(X_test_norm)
0s

✓ [30] reg_pred
0s

array([0.71912284, 1.76401657, 2.70965883, ..., 4.46877017, 1.18751119,
        2.00940251])
```

### Making Predictions

The model has made predictions based on the new data, providing estimated house prices based on the features.

### Calculate Residuals

Calculate the residuals by subtracting the true values from the predicted values:

```
✓ [31] residuals = y_test - reg_pred
0s residuals

20046    -0.242123
3024     -1.306017
15663     2.290351
20484    -0.652926
9814      0.175343
...
15362     0.641254
16623     0.418161
18086     0.531240
2144     -0.464511
3665     -0.494403
Name: Price, Length: 4128, dtype: float64
```

Calculating Residuals

Residuals represent the model's errors. Positive residuals indicate underestimation, while negative residuals indicate overestimation.

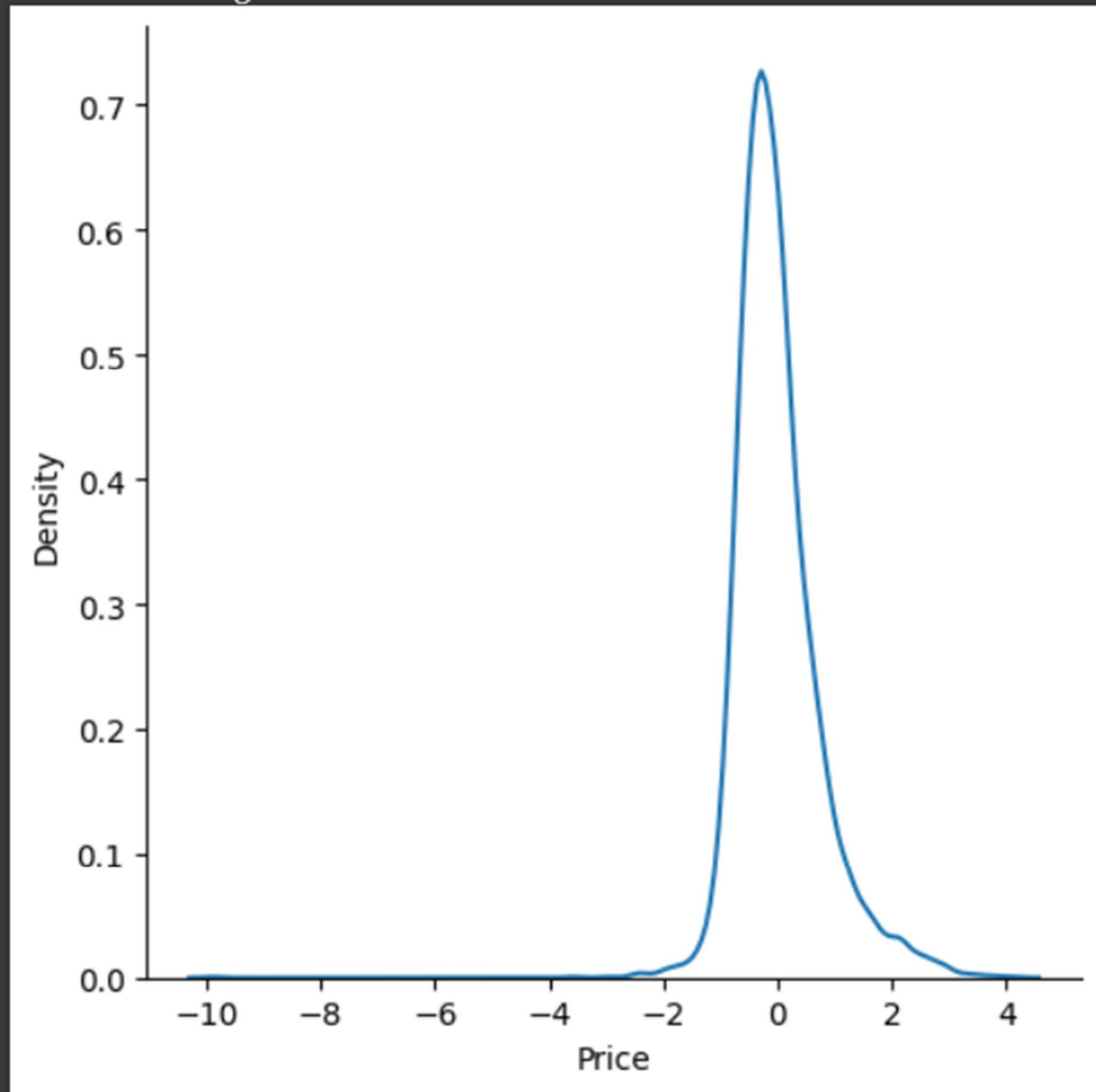
### Visualize Residuals

Create a distribution plot of the residuals to assess their distribution using the `displot` function

```
sns.displot(residuals, kind = 'kde')
```

```
[32] sns.displot(residuals,kind='kde')
```

```
<seaborn.axisgrid.FacetGrid at 0x7a75bc50a2f0>
```



### Plotting Residuals

This plot visually represents the spread and shape of residuals. A symmetric and approximately normal distribution around zero is desirable.