

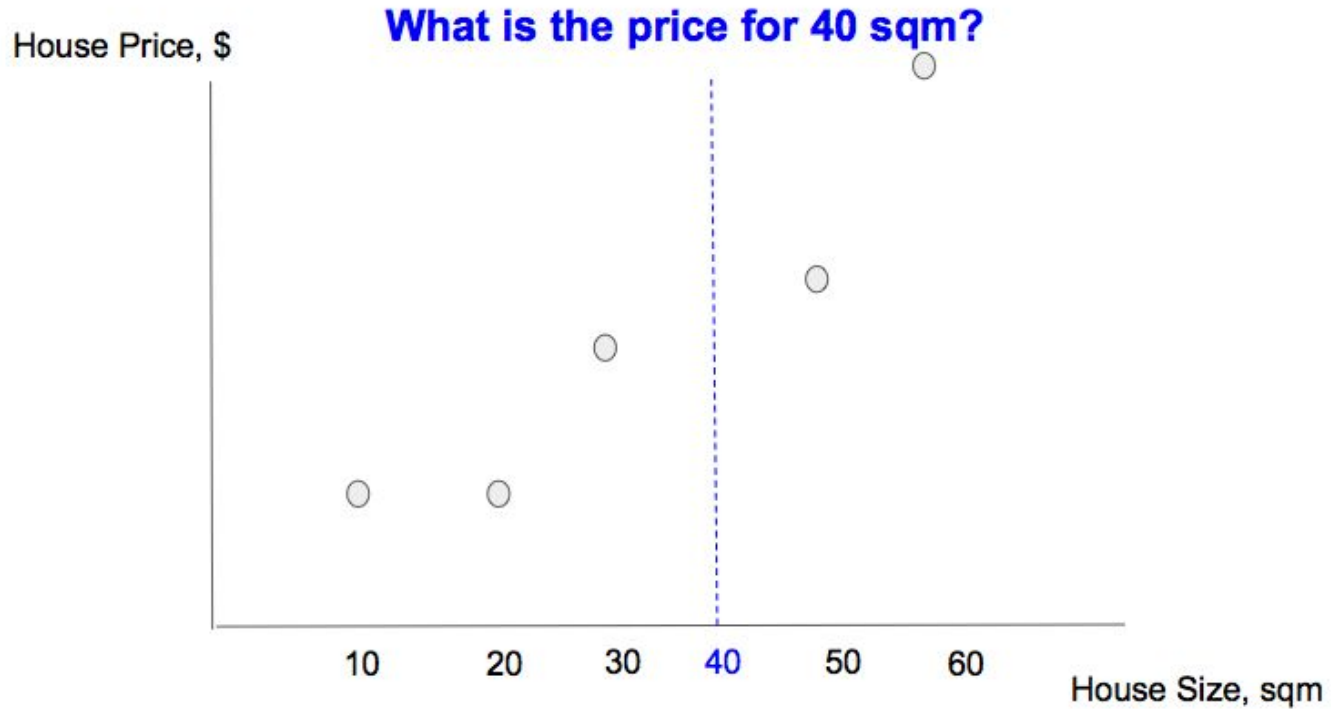
RMSLE: Cost Function

Khor Soon Hin, @neth_6, re:Culture
Katerina Malahova, Tokyo ML Gym

Machine Learning (ML) Lightning Tour

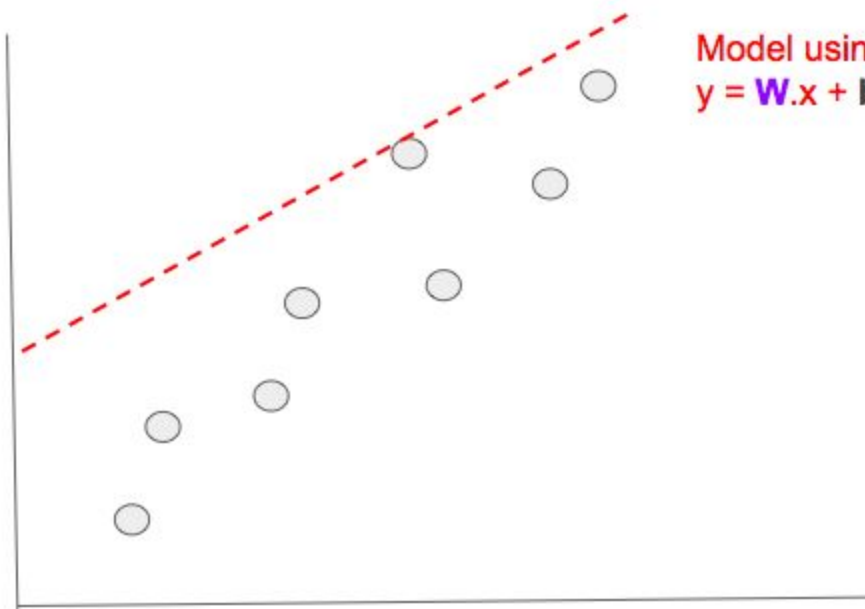
- Linear Regression in ML
- The role of Cost Function

Predicting using ML



Linear Regression

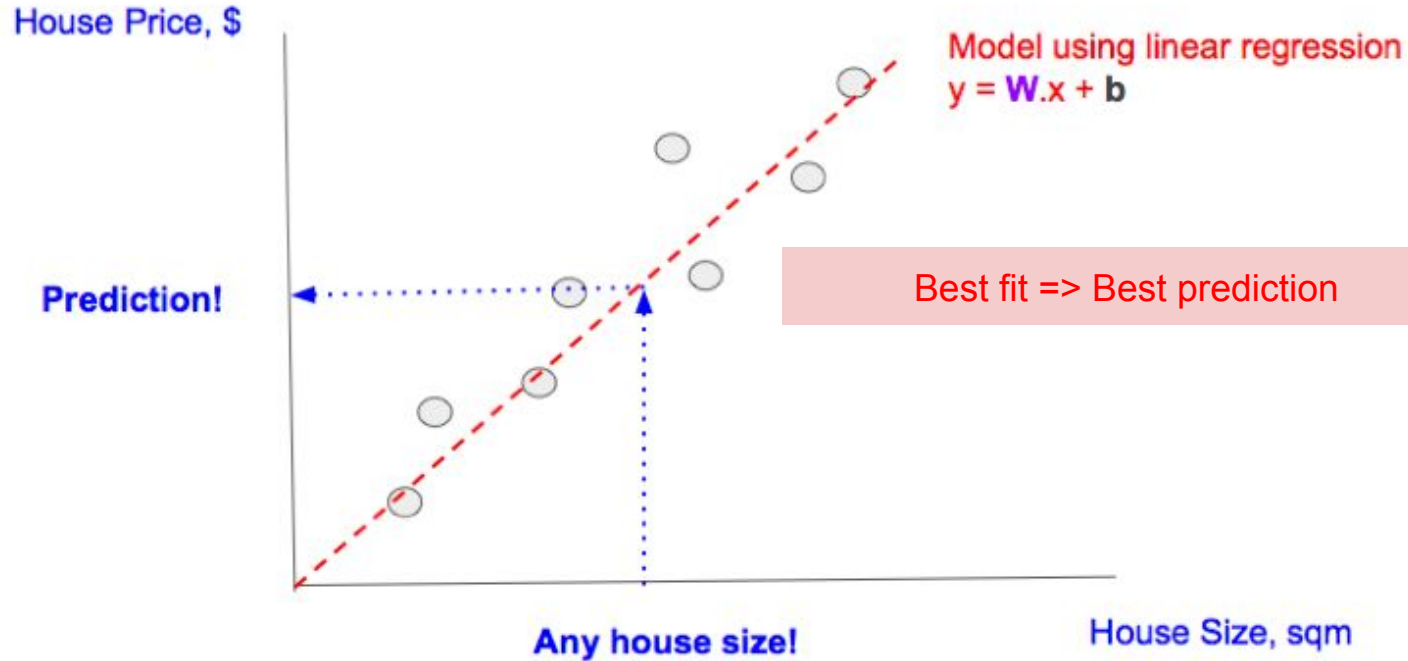
House Price, \$



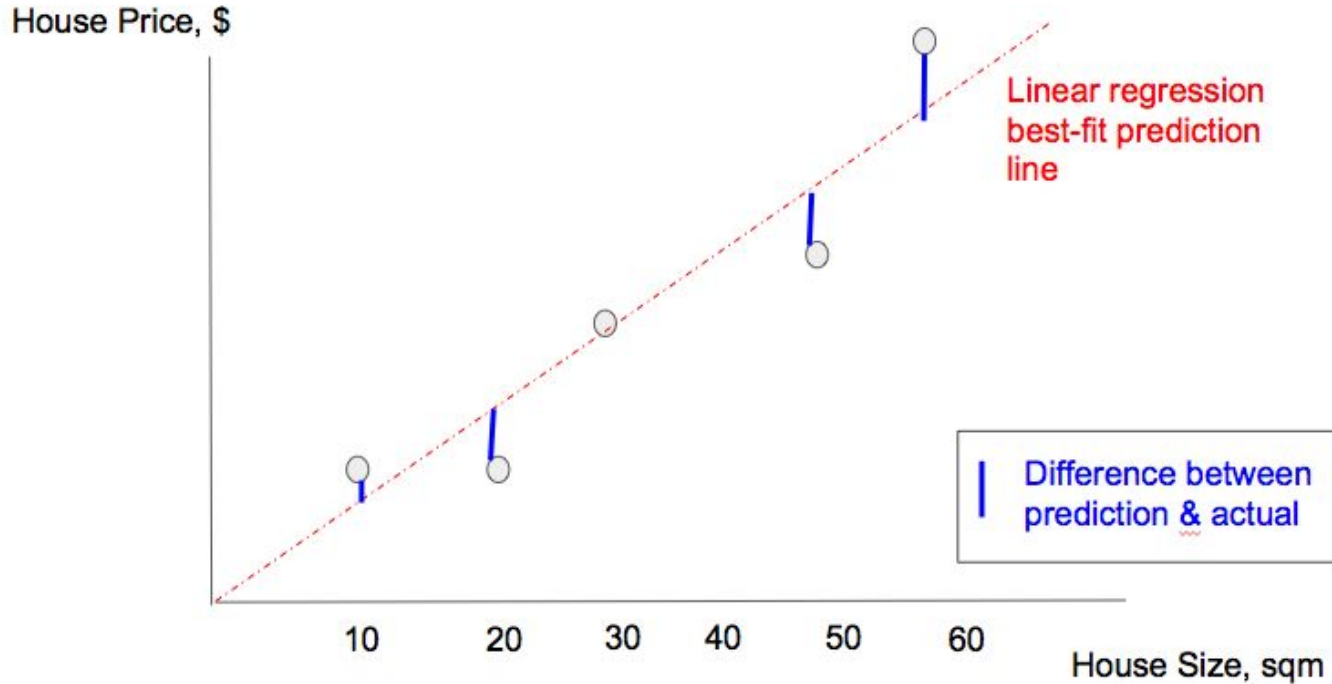
Model using linear regression
 $y = \mathbf{W} \cdot \mathbf{x} + \mathbf{b}$

House Size, sqm

Best Fit



Best Fit Definition: Graphically



Cost Functions

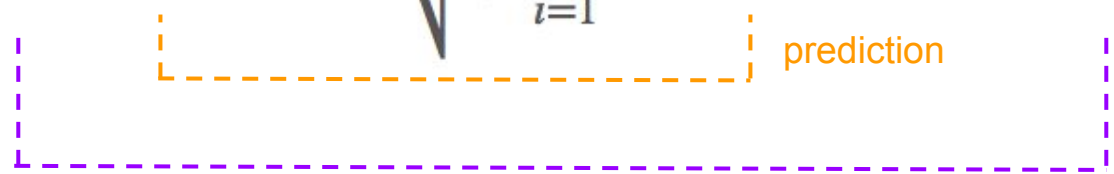
Root Mean Squared Error (RMSE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

Square to ensure always positive

Cost Functions

Root Mean Squared Error (RMSE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$


Root Mean Squared Log Error (RMSLE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (\log(p_i + 1) - \log(a_i + 1))^2}$$

prediction

actual

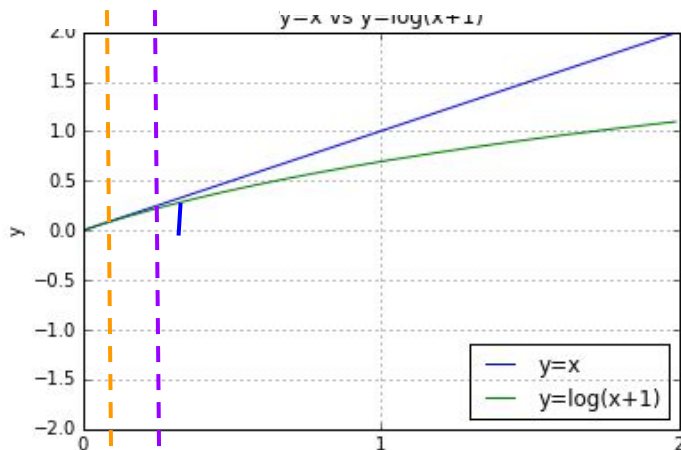
Cost Functions

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Root Mean Squared Log Error (RMSLE)

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When **predicted** and **actual** is small:

For the same **predicted** & **actual**, RMSE & RMSLE is same (the **blue vertical line**)

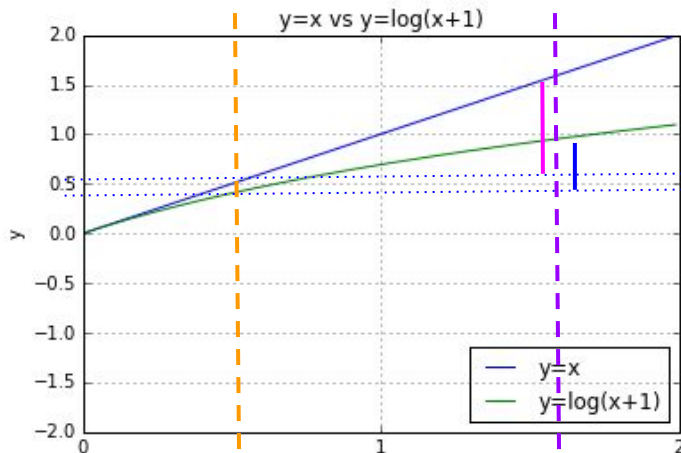
Cost Functions

Root Mean Squared Error (RMSE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

Root Mean Squared Log Error (RMSLE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (\log(p_i + 1) - \log(a_i + 1))^2}$$



If either **predicted** or **actual** is big:

For the same **predicted** & **actual**, $\text{RMSE} > \text{RMSLE}$ is same (the **pink** vs **blue** vertical line)

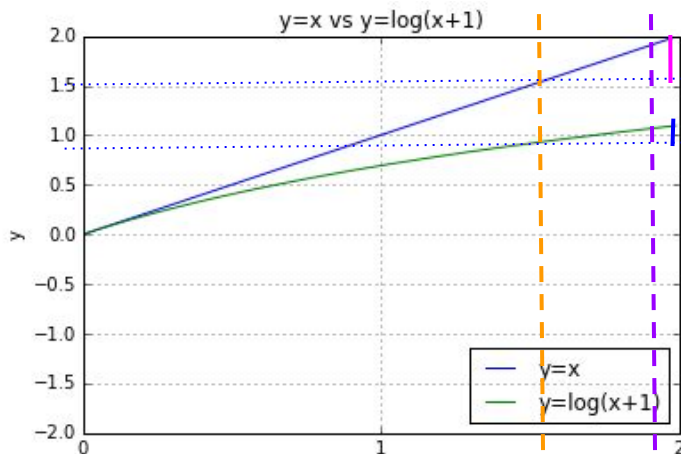
Cost Functions

Root Mean Squared Error (RMSE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

Root Mean Squared Log Error (RMSLE)

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (\log(p_i + 1) - \log(a_i + 1))^2}$$



If both **predicted** and **actual** are big:

For the same **predicted** & **actual**, RMSE > RMSLE is same (the **pink** vs **blue** vertical line)

NOTE: RMSLE is almost negligible

Alternative: RMSLE Intuition

$$\text{RMSLE: } \log(P_i + 1) - \log(A_i + 1) = \log((P_i + 1)/(A_i + 1))$$

Only the percentual differences matter!

For example for $P = 1000$ and $A = 500$ would give you the roughly same error as when $P = 100000$ and $A = 50000$

RMSLE is usually used when you don't want to penalize huge differences in the predicted and true values when both predicted and true values are huge numbers.

Credits to Katerina Malahova for sharing this

RMSLE Usage

- When prediction error for small prediction is undesirable
- Example:
 - Predict inventory required to fulfill all customers
 - When you are small, you have few customers so ensuring each one is satisfied is more important missing out a few customers when you are big
 - Kaggle: <https://www.kaggle.com/wiki/RootMeanSquaredLogarithmicError>

Cost Function

- Purpose:
 - Mathematical definition of best fit
 - ML's goal is to minimize it
- Usage:
 - Used by ML to determine how to tweak model parameters to get best fit
 - Gradient descent