

# Computing Parameters Analytically

## Normal Equation

Assume we have a dataset as follows:

Size ( $ft^2$ )	Beds	Floors	Age	Price (\$10 <sup>3</sup> )
2104	5	1	45	460
1416	3	2	40	232
1534	3	2	30	315
852	2	1	36	178

Here,  $m = 4$

We'll construct matrices  $X$  and  $Y$  as follows:

$$X = \begin{bmatrix} 1 & 2104 & 5 & 1 & 45 \\ 1 & 1416 & 3 & 2 & 40 \\ 1 & 1534 & 3 & 2 & 30 \\ 1 & 852 & 2 & 1 & 36 \end{bmatrix}_{[m \times (n+1)]} \quad y = \begin{bmatrix} 460 \\ 232 \\ 315 \\ 178 \end{bmatrix}_{[m \times 1]}$$

Add an extra column  $x_0$  with all 1's to the matrix  $X$ .

Now, we'll compute  $\theta = (X^T \cdot X)^{-1} \cdot X^T \cdot y$

For using this in Octave,

```
pinv(X' * X) * X' * y
```

**Note:** Feature scaling is not necessary in case we're using Normal Equation

Gradient Descent	Normal Equation
Need to choose the learning rate $\alpha$	No need to choose the learning rate $\alpha$
Needs many iteration	Doesn't need any iteration
Requires less computation	Requires huge computation of $O(n^3)$ since calculation of $(X^T \cdot X)^{-1}$ is needed
Works well even when $n$ is large ( $n$ is the number of features)	Becomes slow if $n$ is large

When to use Normal Equation?

We can comfortably use normal equation **when the value of 'n' (features) is less than 10,000**. More than that, modern computers will take a lot of time for computation.