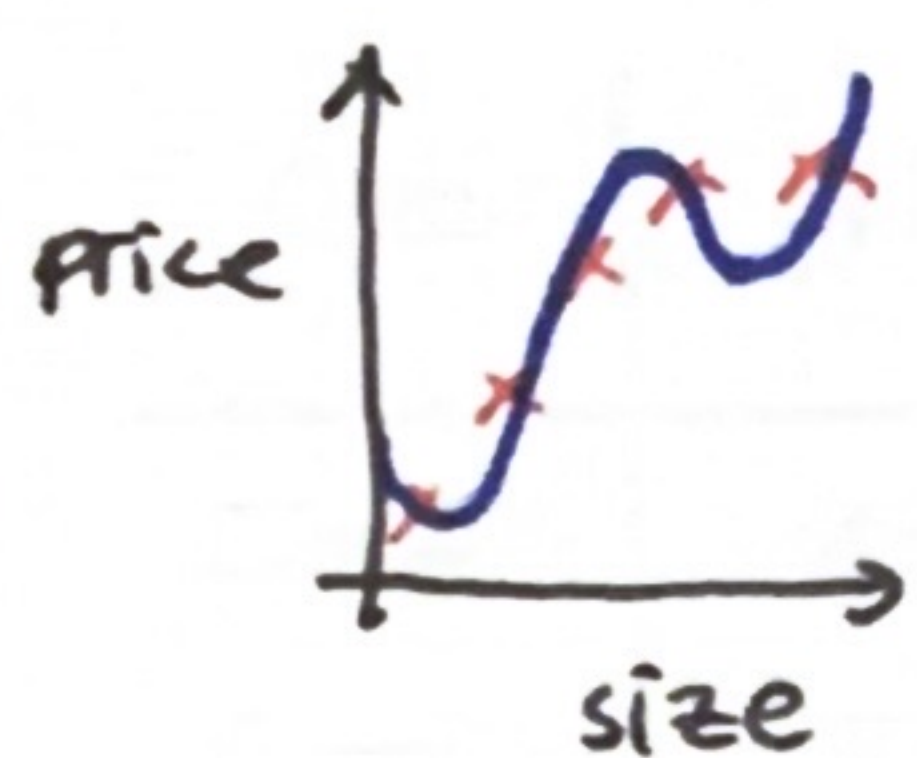


## < Model Selection and Training/Validation/Test sets >



⇒ Once parameters  $\vec{w}, b$  are fit to the training set, the training error  $J_{\text{train}}(\vec{w}, b)$  is likely lower than the actual generalization error.

$$f_{\vec{w}, b}(\vec{x}) = w_1 x_1 + w_2 x_2^2 + w_3 x_3^3 + w_4 x_4^4 + b$$

⇒  $J_{\text{train}}(\vec{w}, b)$  : not good indicator of how well the model will likely do on new data

⇒  $J_{\text{test}}(\vec{w}, b)$  is better estimate of how well the model will generalize to new data than  $J_{\text{train}}(\vec{w}, b)$

\* Use test set to choose a model

$d$  = degree of polynomial

$$d=1 : f_{\vec{w}, b}(\vec{x}) = w_1 x_1 + b$$

$$d=2 : f_{\vec{w}, b}(\vec{x}) = w_1 x_1 + w_2 x_2^2 + b$$

$$d=3 : f_{\vec{w}, b}(\vec{x}) = w_1 x_1 + w_2 x_2^2 + w_3 x_3^3 + b$$

⋮

$$d=10 : f_{\vec{w}, b}(\vec{x}) = w_1 x_1 + w_2 x_2^2 + \dots + w_{10} x_{10}^{10} + b \rightarrow \vec{w}^{(10)}, b^{(10)} \xrightarrow{\text{test}} J_{\text{test}}(\vec{w}^{(10)}, b^{(10)})$$

↓

if find  $J_{\text{test}}(\vec{w}^{(5)}, b^{(5)})$  is the lowest

⇒ 이렇게 계산된 일련의  $J_{\text{test}}()$  중  $d=5$  일 때의 test error가 가장 작아서

$w_1 x_1 + w_2 x_2^2 + \dots + w_5 x_5^5 + b$  을 모델로 선택했다고 하면,

이때  $J_{\text{test}}(\vec{w}^{(5)}, b^{(5)})$  가 model의 generalization을 가장 잘 반영한 것일까?

⇒ 사실  $J_{\text{test}}(\vec{w}^{(5)}, b^{(5)})$  는 generalization error의 optimistic estimate인 가능성이 높음  
(= lower than the actual generalization error)

∴ An extra parameter  $d$  (degree of polynomial) was chosen using the test set.

↳ test set에 fit되었기 때문

∴ 주어진 데이터를 단순히 training/test로 나누는 것이 아니라  $d$ 를 fit하기 위한 set 역시 만들어야 함.