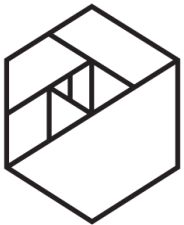


Model Selection II



METIS

datascopes

Training and test sets

	Date	Title	Budget	DomesticTotalGross	Director	Rating	Runtime
0	2013-11-22	The Hunger Games: Catching Fire	130000000	424668047	Francis Lawrence	PG-13	146
1	2013-05-03	Iron Man 3	200000000	409013994	Shane Black	PG-13	129
2	2013-11-22	Frozen	150000000	400738009	Chris BuckJennifer Lee	PG	108
3	2013-07-03	Despicable Me 2	76000000	368061265	Pierre CoffinChris Renaud	PG	98
4	2013-06-14	Man of Steel	225000000	291045518	Zack Snyder	PG-13	143
5	2013-10-04	Gravity	100000000	274092705	Alfonso Cuaron	PG-13	91
6	2013-06-21	Monsters University	NaN	268492764	Dan Scanlon	G	107
7	2013-12-13	The Hobbit: The Desolation of Smaug	NaN	258366855	Peter Jackson	PG-13	161
8	2013-05-24	Fast & Furious 6	160000000	238679850	Justin Lin	PG-13	130
9	2013-03-08	Oz The Great and Powerful	215000000	234911825	Sam Raimi	PG	127
10	2013-05-16	Star Trek Into Darkness	190000000	228778661	J.J. Abrams	PG-13	123
11	2013-11-08	Thor: The Dark World	170000000	206362140	Alan Taylor	PG-13	120
12	2013-06-21	World War Z	190000000	202359711	Marc Forster	PG-13	116
13	2013-03-22	The Croods	135000000	187168425	Kirk De MiccoChris Sanders	PG	98
14	2013-06-28	The Heat	43000000	159582188	Paul Feig	R	117
15	2013-08-07	We're the Millers	37000000	150394119	Rawson Marshall Thurber	R	110
16	2013-12-13	American Hustle	40000000	150117807	David O. Russell	R	138
17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

Training and test sets

	Date	Title	Budget	DomesticTotalGross	Director	Rating	Runtime
0	2013-11-22	The Hunger Games: Catching Fire	130000000	424668047	Francis Lawrence	PG-13	146
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2	2013-11-22	Frozen	150000000	400738009	Chris BuckJennifer Lee	PG	108
3	2013-07-03	Despicable Me 2	76000000	368061265	Pierre CoffinChris Renaud	PG	98
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7	2013-12-13	The Hobbit: The Desolation of Smaug	NaN	258366855	Peter Jackson	PG-13	161
8	2013-05-24	Fast & Furious 6	160000000	238679850	Justin Lin	PG-13	130
9	2013-03-08	Oz The Great and Powerful	215000000	234911825	Sam Raimi	PG	127
10	2013-05-16	Star Trek Into Darkness	190000000	228778661	J.J. Abrams	PG-13	123
11	2013-11-08	Thor: The Dark World	170000000	206362140	Alan Taylor	PG-13	120
12	2013-06-21	World War Z	190000000	202359711	Marc Forster	PG-13	116
13	2013-03-22	The Croods	135000000	187168425	Kirk De MiccoChris Sanders	PG	98
14	2013-06-28	The Heat	43000000	159582188	Paul Feig	R	117
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17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

Training
set

Test
set

Training and test sets

Training set

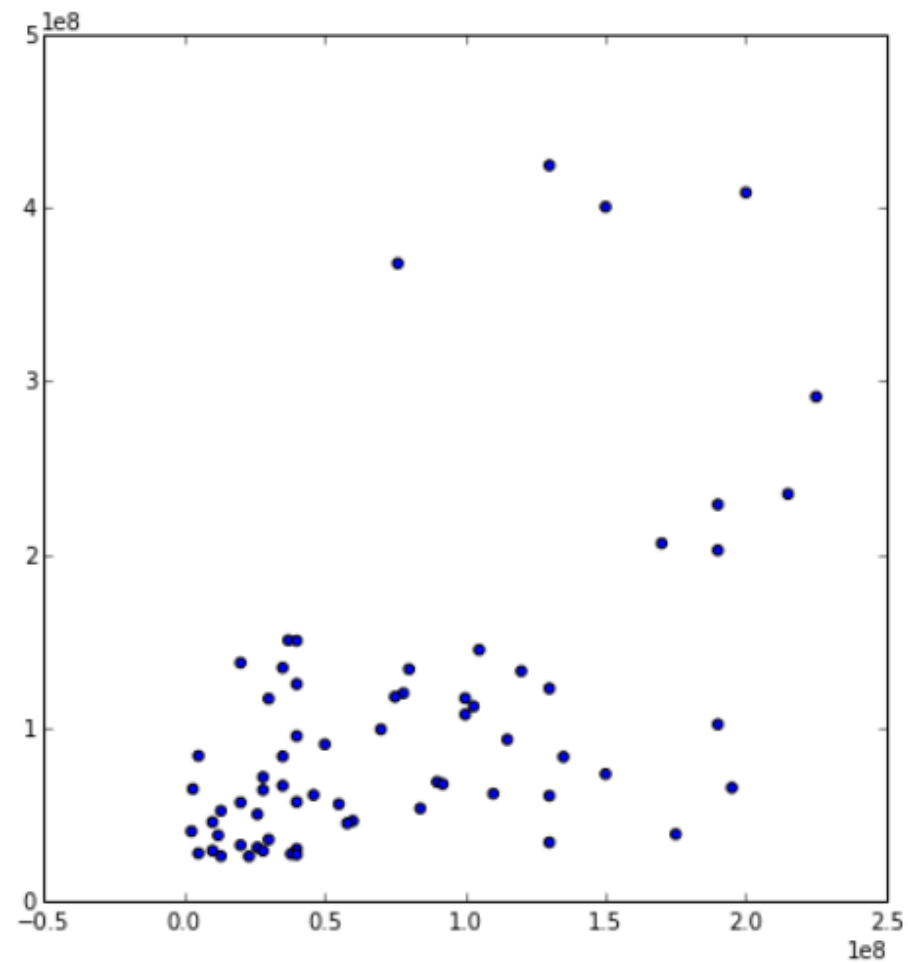
fit the model

Test set

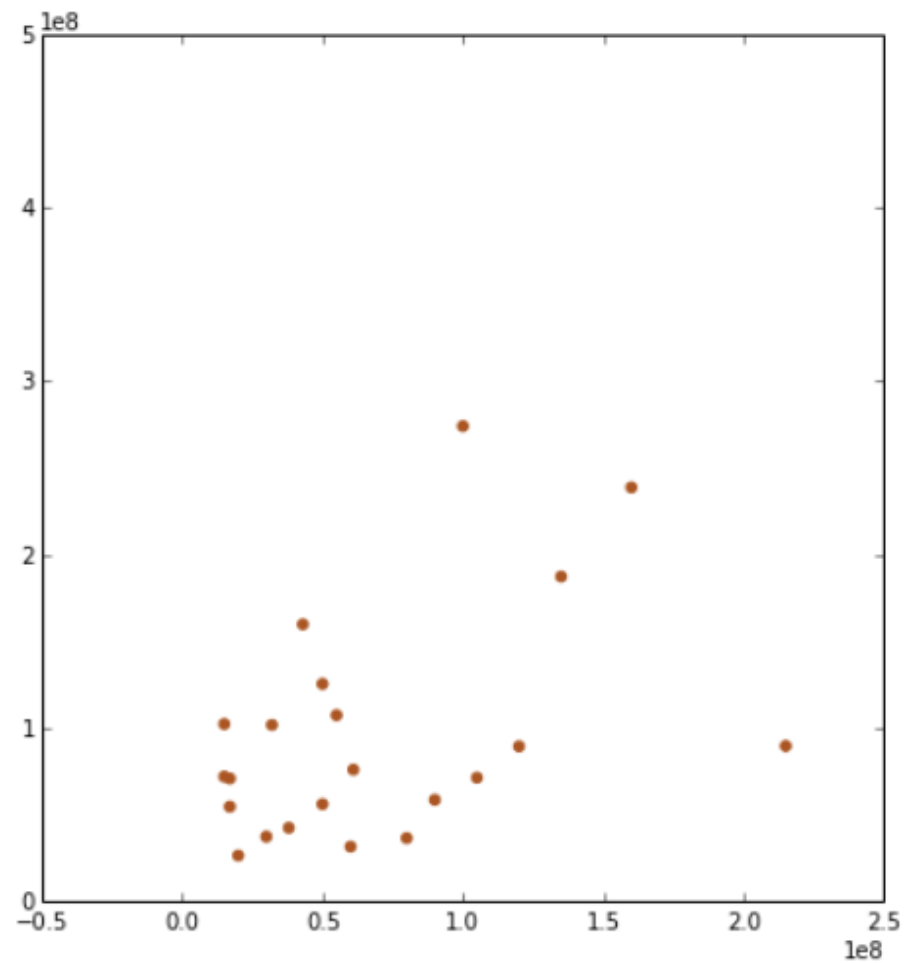
measure performance

- predict y with model
- compare with actual y
- measure error

Training set

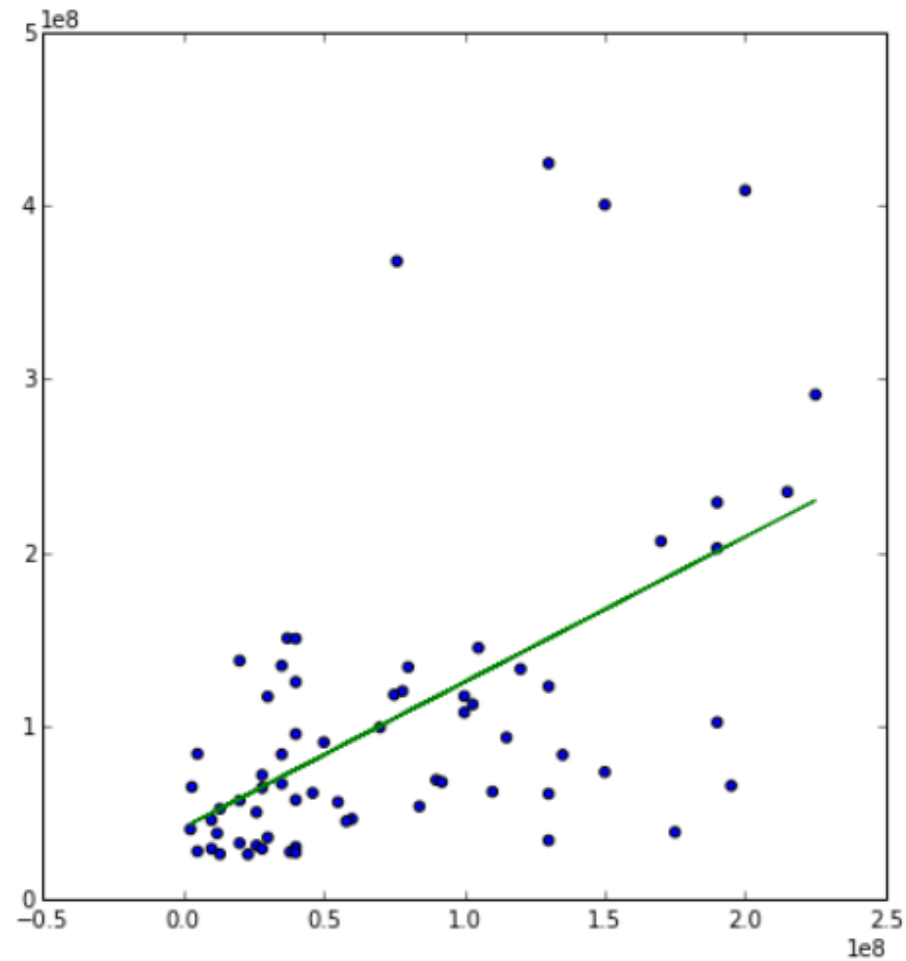


Test set

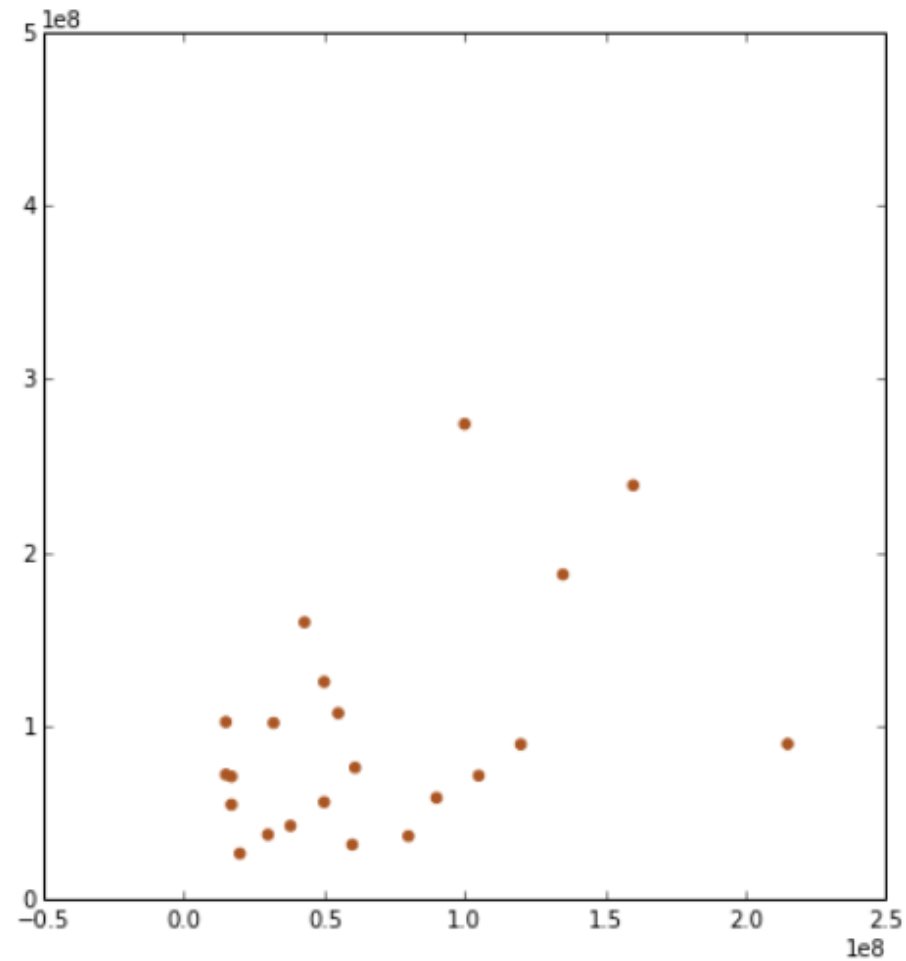


Training set

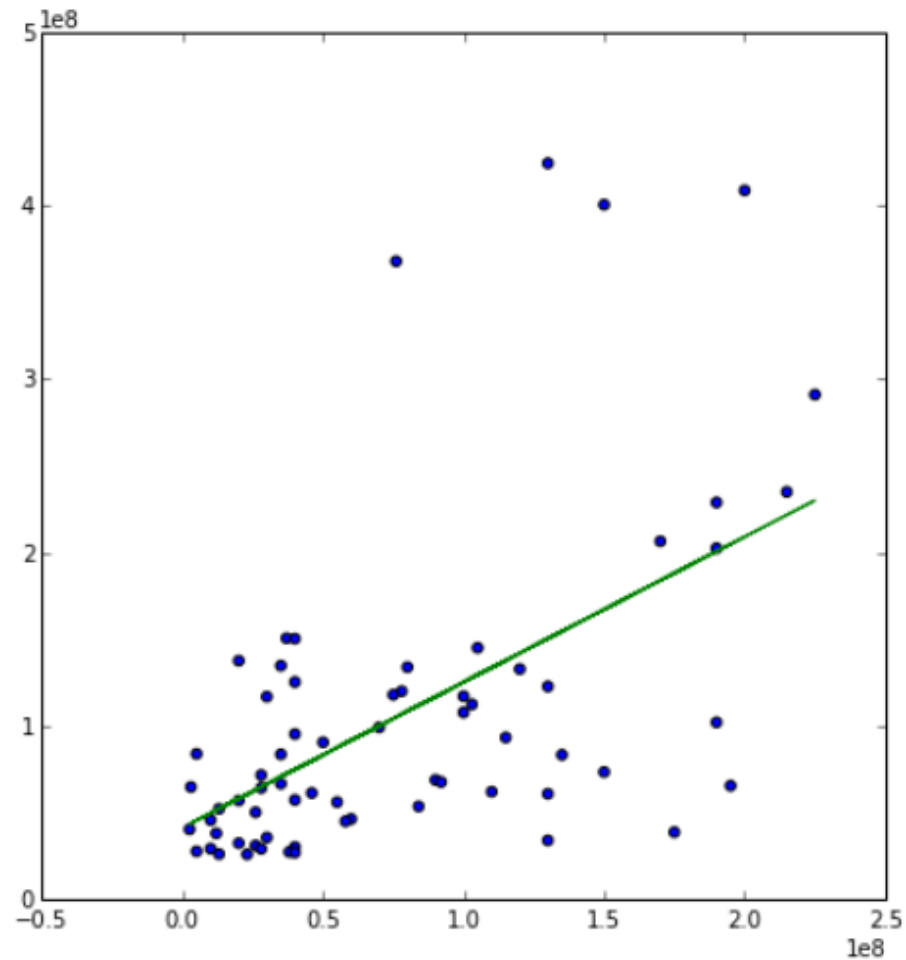
Fit the model



Test set

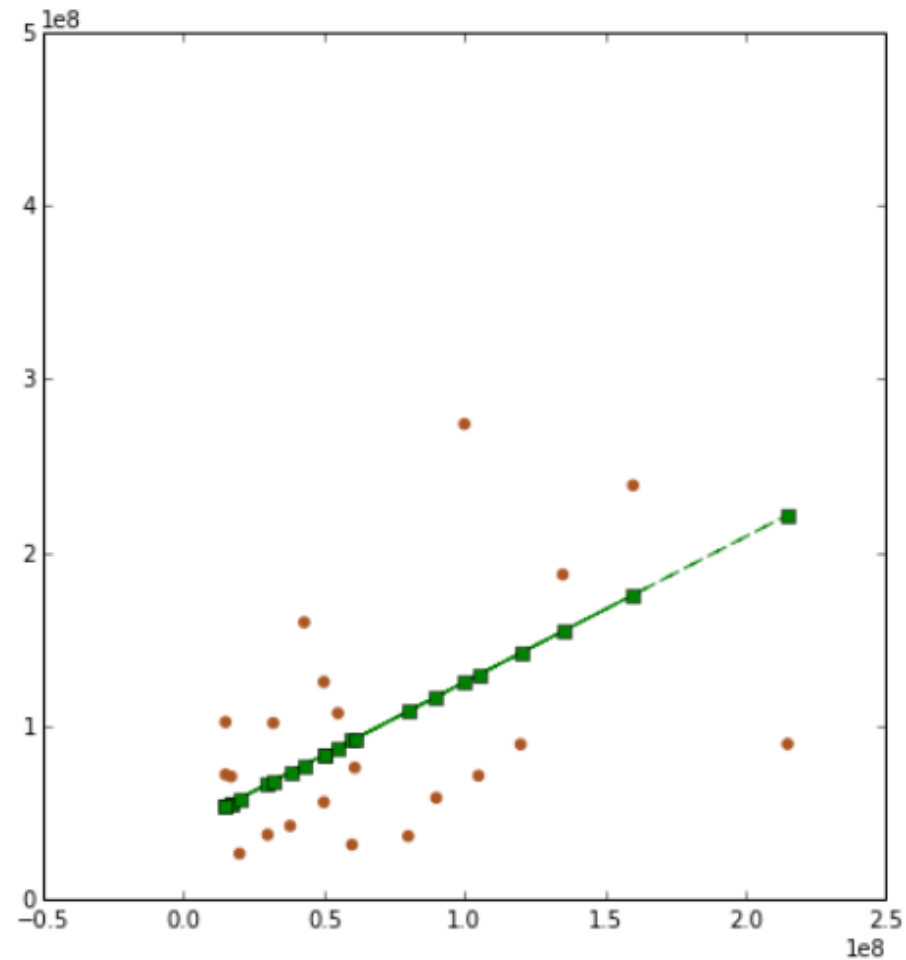


Training set

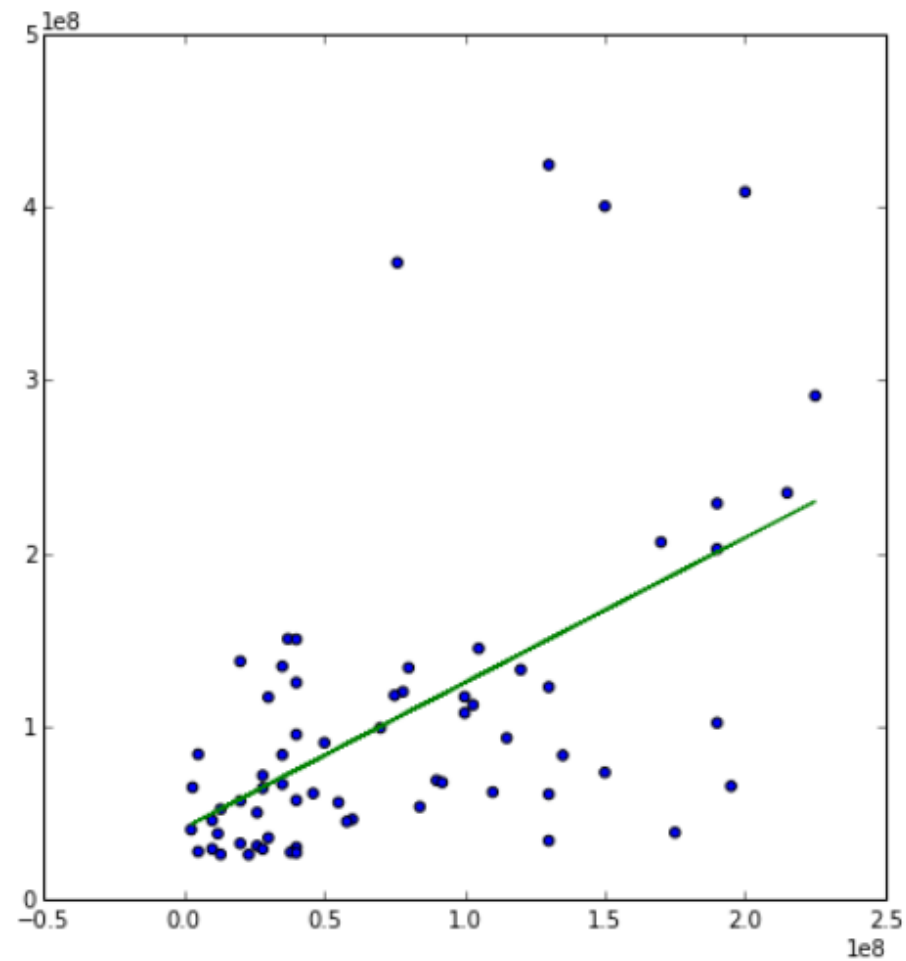


Test set

Use the model
to predict y from x

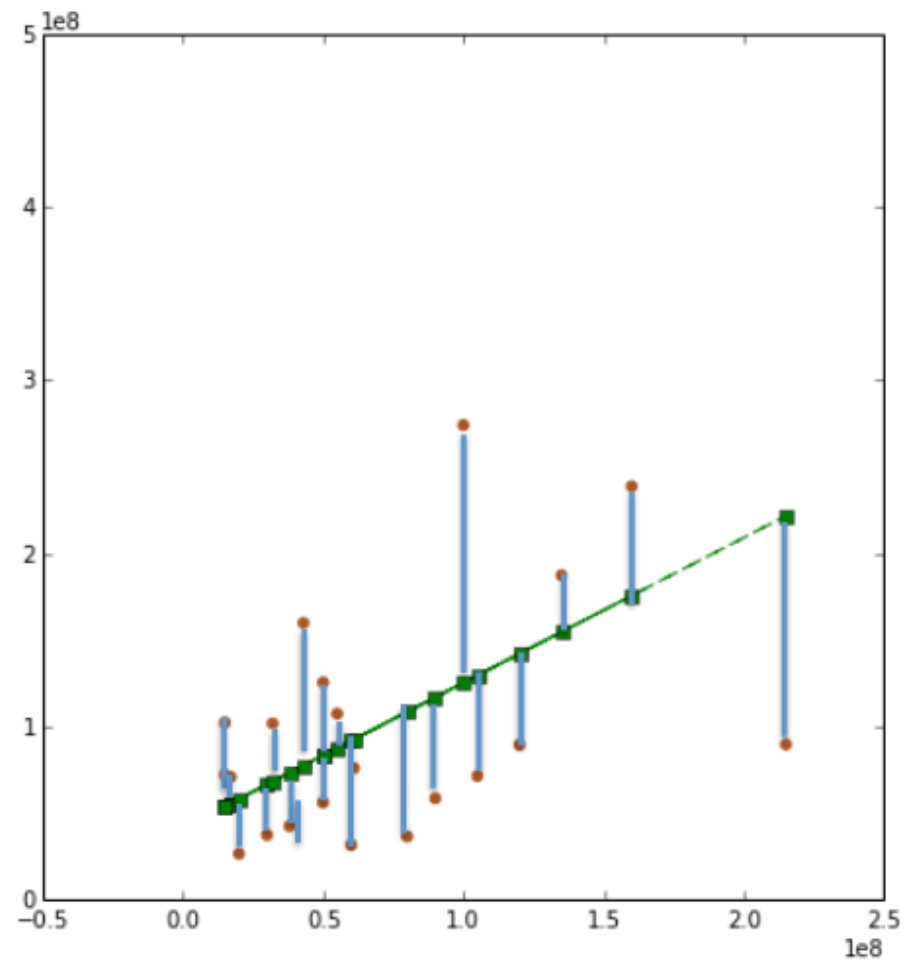


Training set

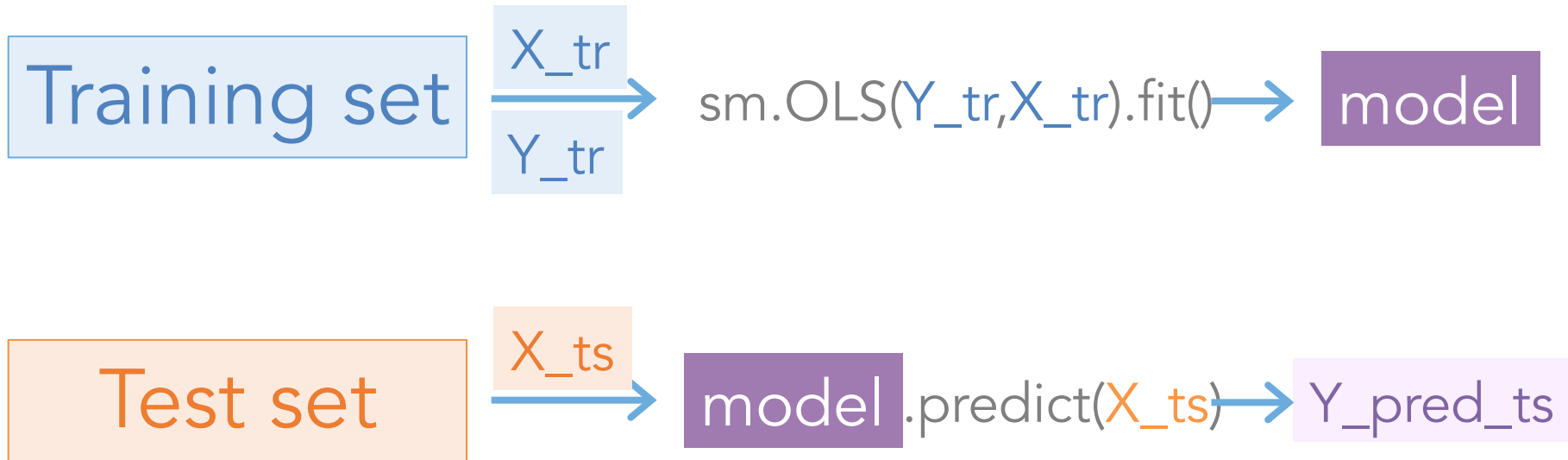


Test set

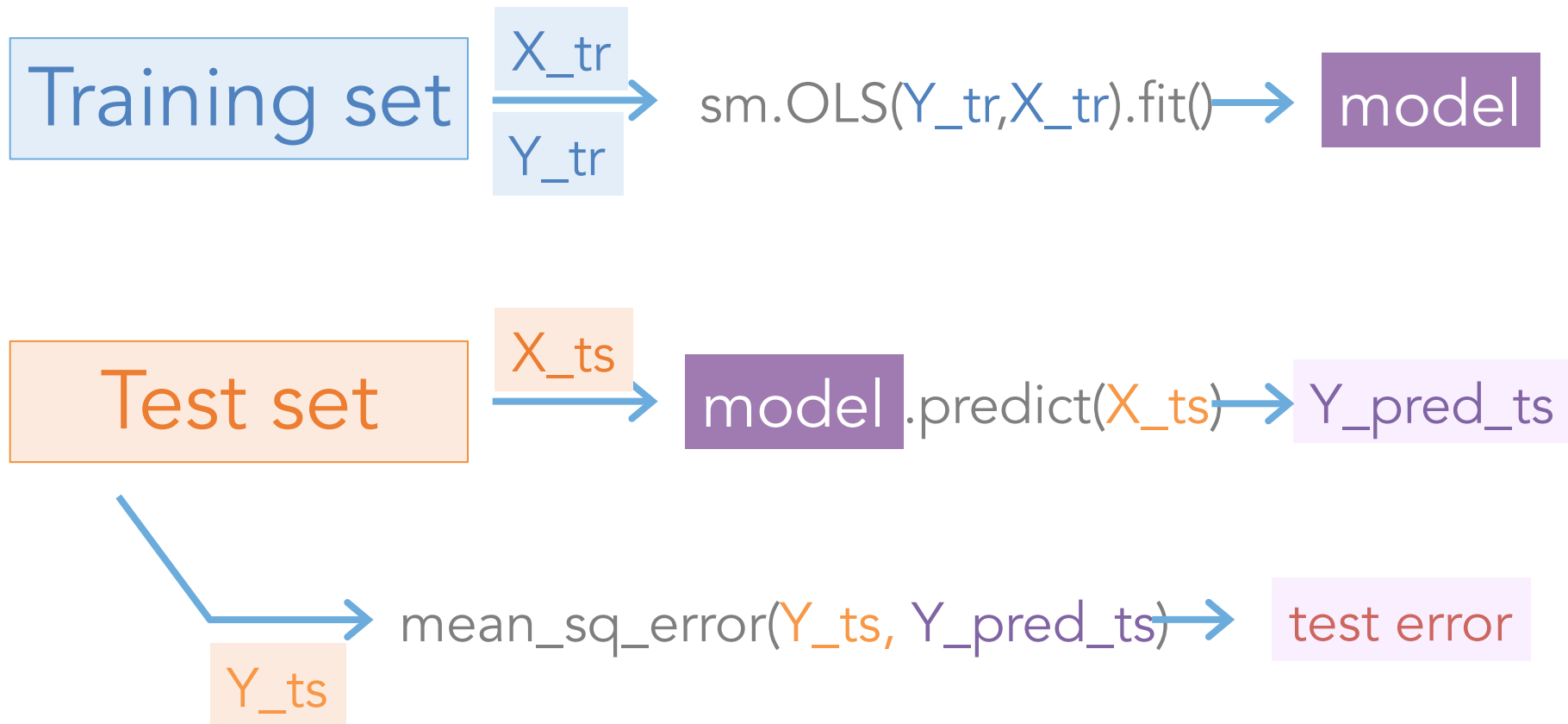
Measure the
error



Training and test sets



Training and test sets



Cross validation

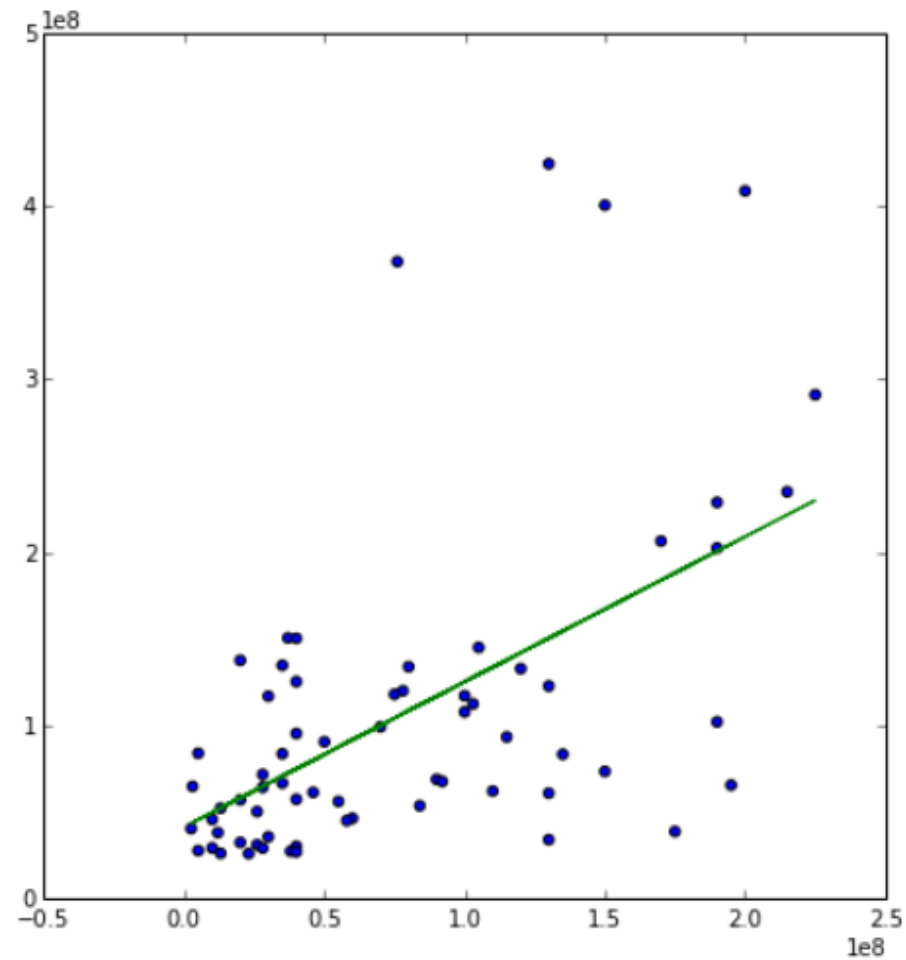
What's better than a single test set?

	Date	Title	Budget	DomesticTotalGross	Director	Rating	Runtime
0	2013-11-22	The Hunger Games: Catching Fire	130000000	424668047	Francis Lawrence	PG-13	146
1	2013-05-03	Iron Man 3	200000000	409013994	Shane Black	PG-13	129
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5	2013-10-04	Gravity	100000000	274092705	Alfonso Cuaron	PG-13	91
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7	2013-12-13	The Hobbit: The Desolation of Smaug	NaN	258366855	Peter Jackson	PG-13	161
8	2013-05-24	Fast & Furious 6	160000000	238679850	Justin Lin	PG-13	130
9	2013-03-08	Oz The Great and Powerful	215000000	234911825	Sam Raimi	PG	127
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11	2013-11-08	Thor: The Dark World	170000000	206362140	Alan Taylor	PG-13	120
12	2013-06-21	World War Z	190000000	202359711	Marc Forster	PG-13	116
13	2013-03-22	The Croods	135000000	187168425	Kirk De MiccoChris Sanders	PG	98
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15	2013-08-07	We're the Millers	37000000	150394119	Rawson Marshall Thurber	R	110
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17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

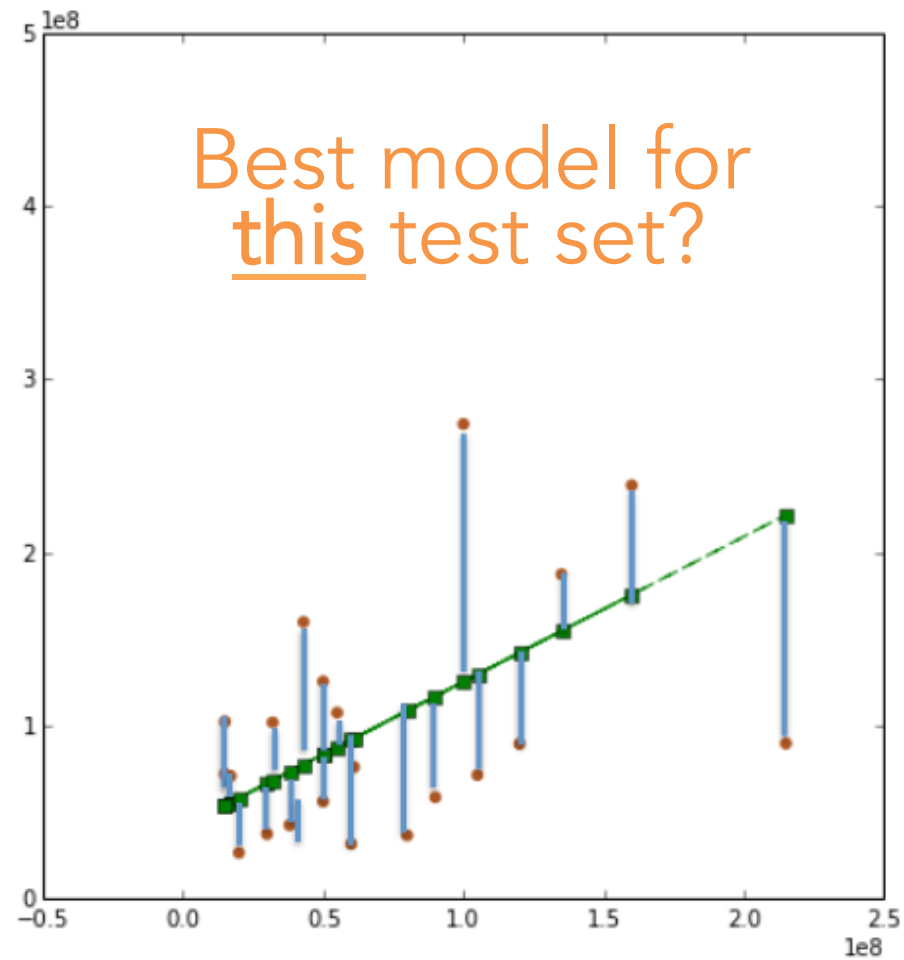
Training
set

Test
set

Training set



Test set



Cross validation

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0	2013-11-22	The Hunger Games: Catching Fire	130000000	424668047	Francis Lawrence	PG-13	146
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16	2013-12-13	American Hustle	40000000	150117807	David O. Russell	R	138
17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

Training
set

Test
set

Cross validation

	Date	Title	Budget	DomesticTotalGross	Director	Rating	Runtime
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16	2013-12-13	American Hustle	40000000	150117807	David O. Russell	R	138
17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

Training
set 2

Test
set 2

Cross validation

	Date	Title	Budget	DomesticTotalGross	Director	Rating	Runtime
0	2013-11-22	The Hunger Games: Catching Fire	130000000	424668047	Francis Lawrence	PG-13	146
1	2013-05-03	Iron Man 3	200000000	409013994	Shane Black	PG-13	129
2	2013-11-22	Frozen	150000000	400738009	Chris BuckJennifer Lee	PG	108
3	2013-07-03	Despicable Me 2	76000000	368061265	Pierre CoffinChris Renaud	PG	98
4	2013-06-14	Man of Steel	225000000	291045518	Zack Snyder	PG-13	143
5	2013-10-04	Gravity	100000000	274092705	Alfonso Cuaron	PG-13	91
6	2013-06-21	Monsters University	NaN	268492764	Dan Scanlon	G	107
7	2013-12-13	The Hobbit: The Desolation of Smaug	NaN	258366855	Peter Jackson	PG-13	161
8	2013-05-24	Fast & Furious 6	160000000	238679850	Justin Lin	PG-13	130
9	2013-03-08	Oz The Great and Powerful	215000000	234911825	Sam Raimi	PG	127
10	2013-05-16	Star Trek Into Darkness	190000000	228778661	J.J. Abrams	PG-13	123
11	2013-11-08	Thor: The Dark World	170000000	206362140	Alan Taylor	PG-13	120
12	2013-06-21	World War Z	190000000	202359711	Marc Forster	PG-13	116
13	2013-03-22	The Croods	135000000	187168425	Kirk De MiccoChris Sanders	PG	98
14	2013-06-28	The Heat	43000000	159582188	Paul Feig	R	117
15	2013-08-07	We're the Millers	37000000	150394119	Rawson Marshall Thurber	R	110
16	2013-12-13	American Hustle	40000000	150117807	David O. Russell	R	138
17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

Test
set 3

Training
Set 3

Cross validation

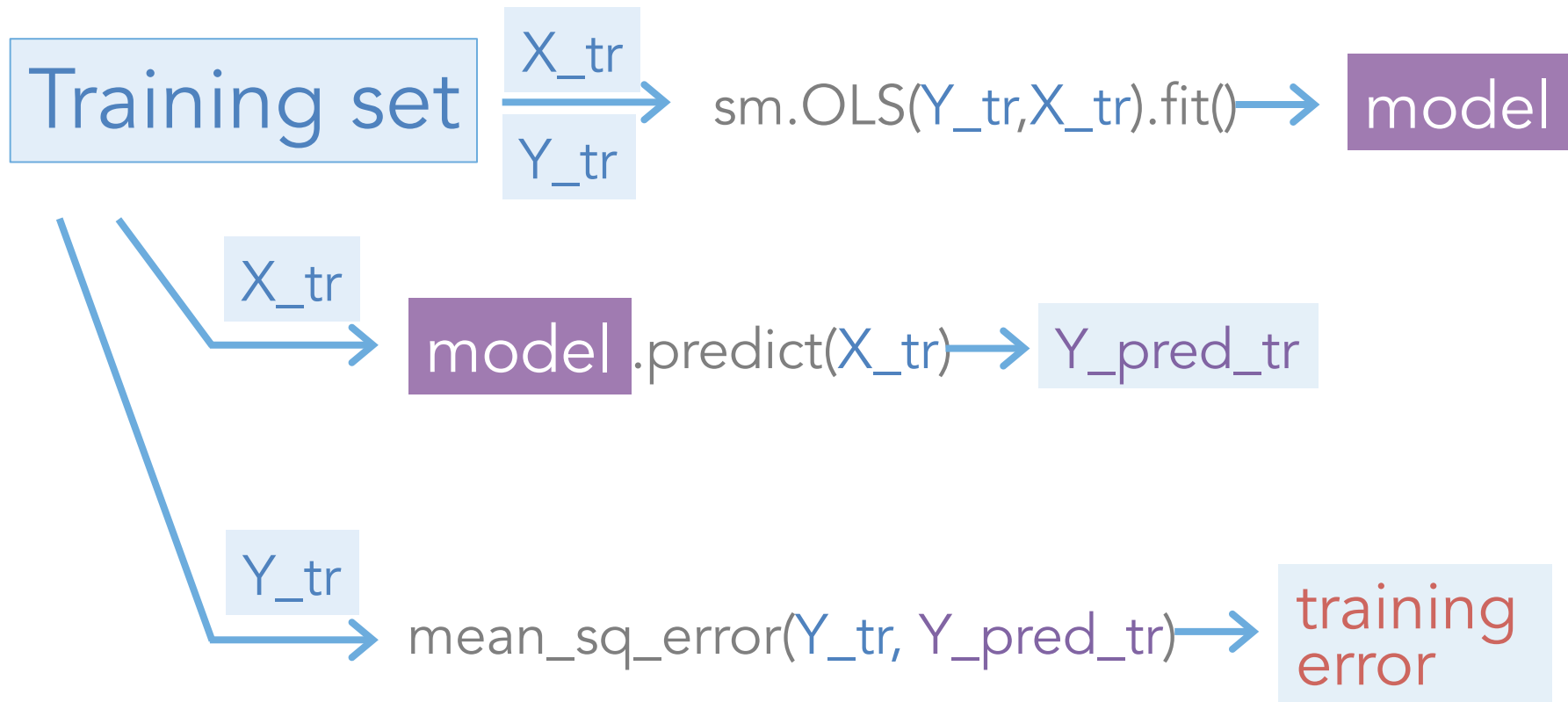
	Date	Title	Budget	DomesticTotalGross	Director	Rating	Runtime
0	2013-11-22	The Hunger Games: Catching Fire	130000000	424668047	Francis Lawrence	PG-13	146
1	2013-05-03	Iron Man 3	200000000	409013994	Shane Black	PG-13	129
2	2013-11-22	Frozen	150000000	400738009	Chris BuckJennifer Lee	PG	108
3	2013-07-03	Despicable Me 2	76000000	368061265	Pierre CoffinChris Renaud	PG	98
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8	2013-05-24	Fast & Furious 6	160000000	238679850	Justin Lin	PG-13	130
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17	2013-05-10	The Great Gatsby	105000000	144840419	Baz Luhrmann	PG-13	143

Test
set 4

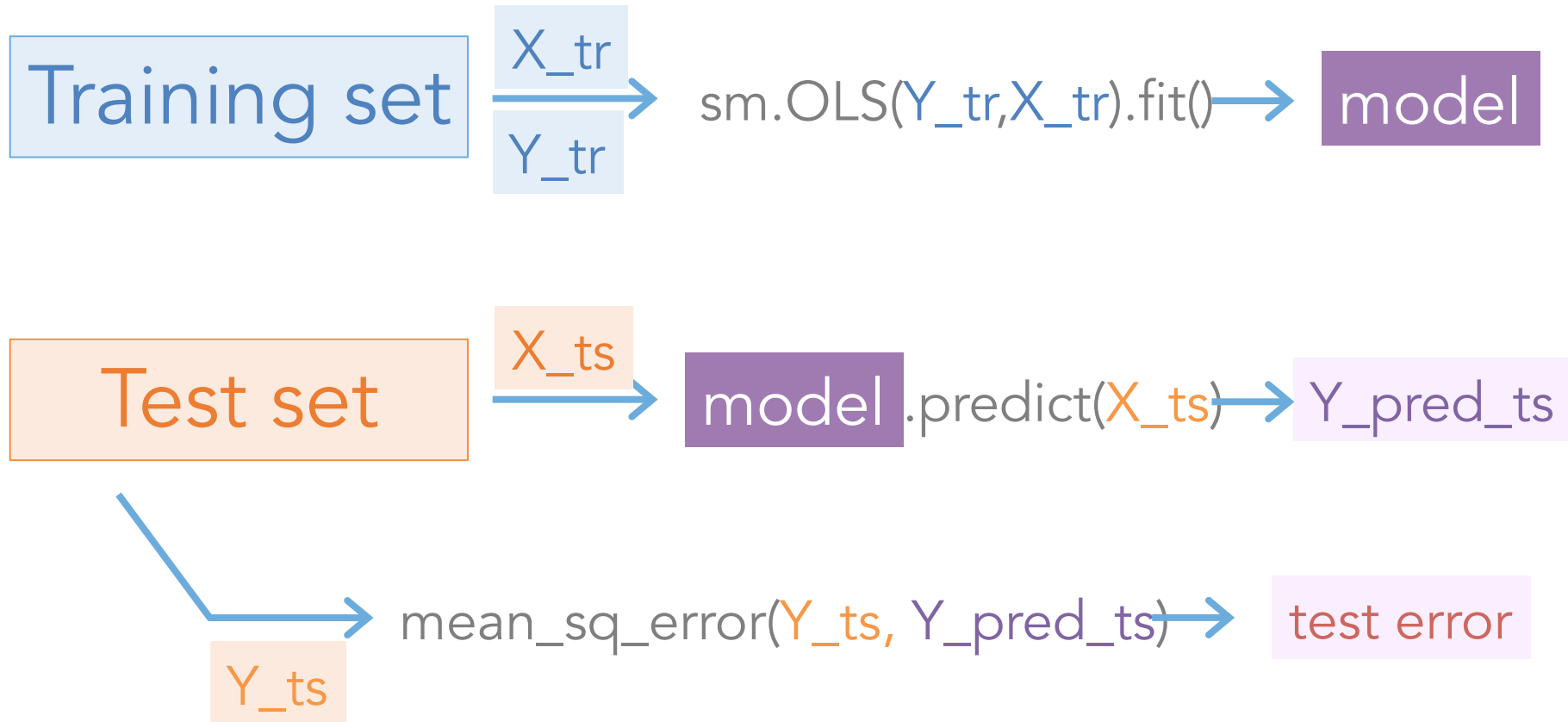
Training
Set 4

Diagnostics with training and test errors

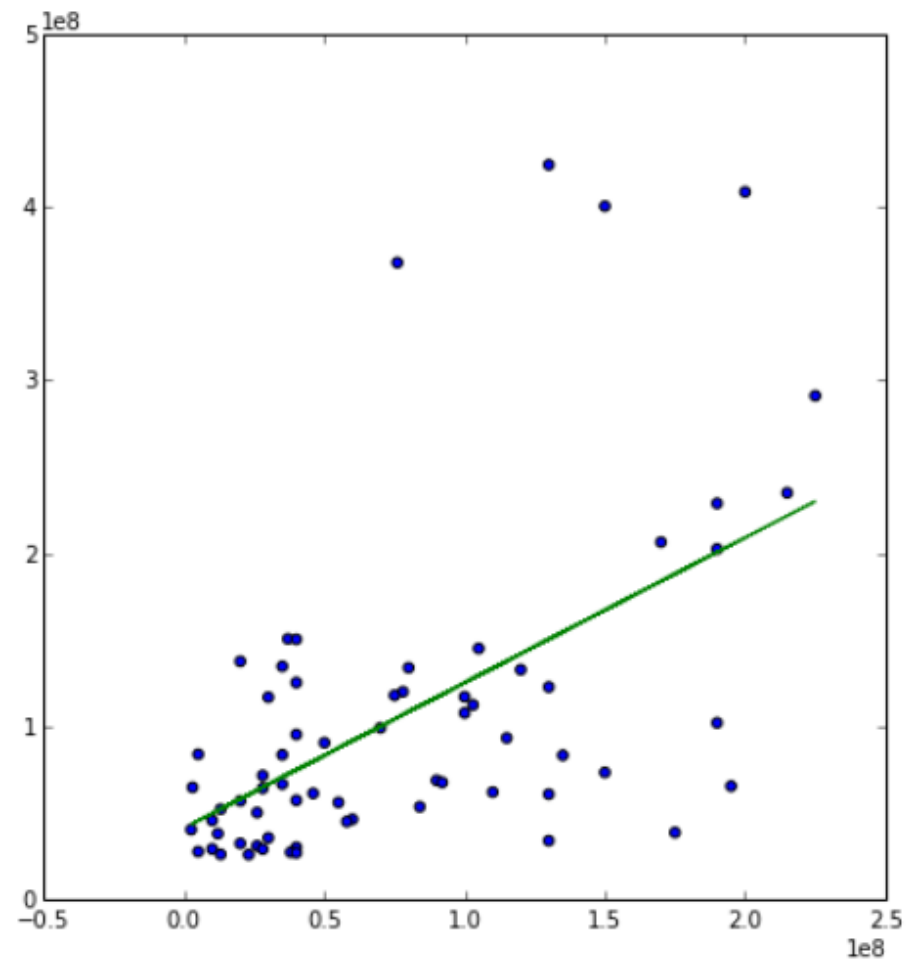
Calculating Training error



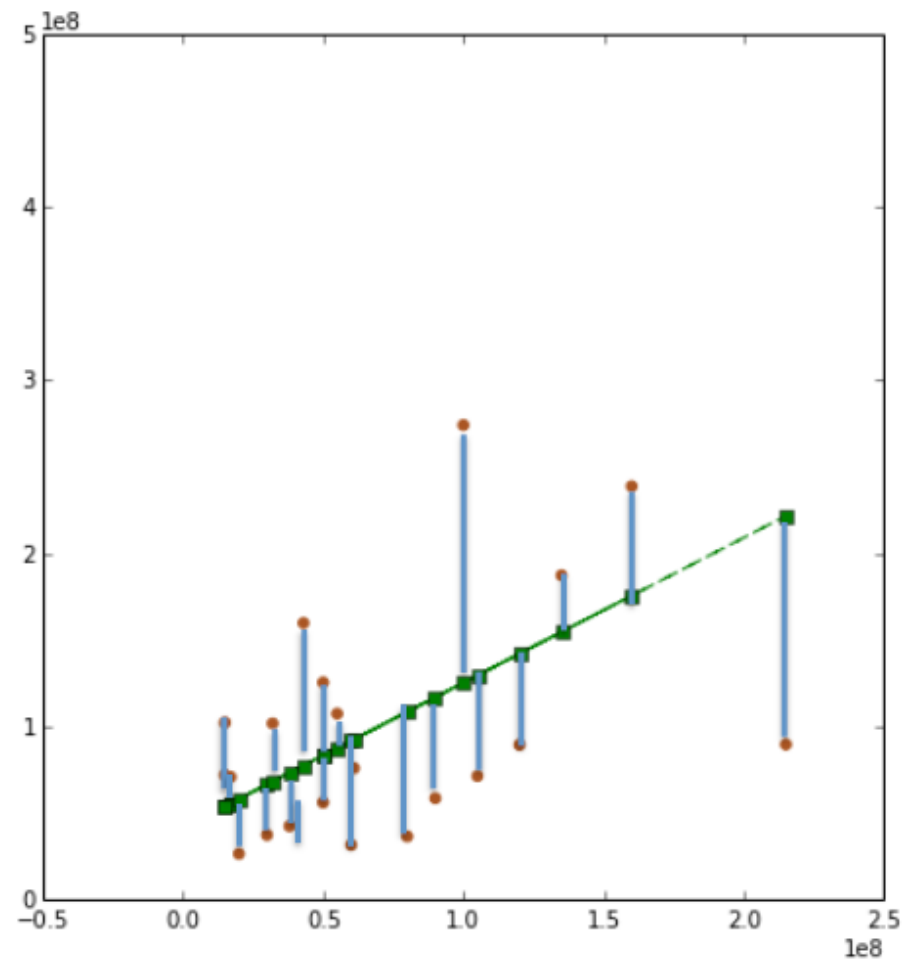
Calculating test error



Training set

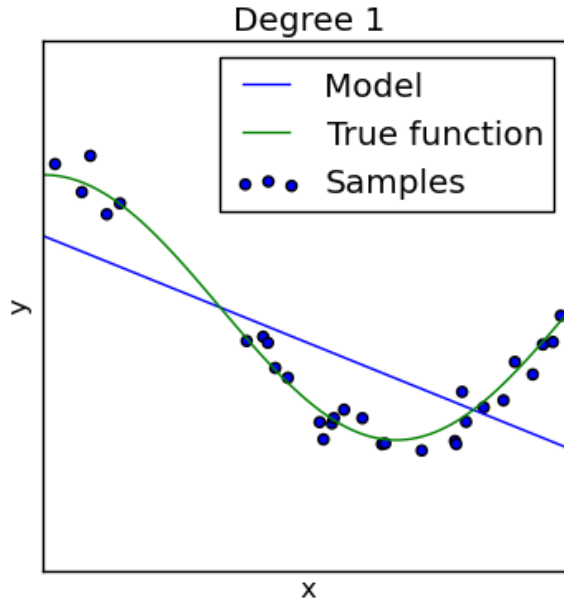


Test set

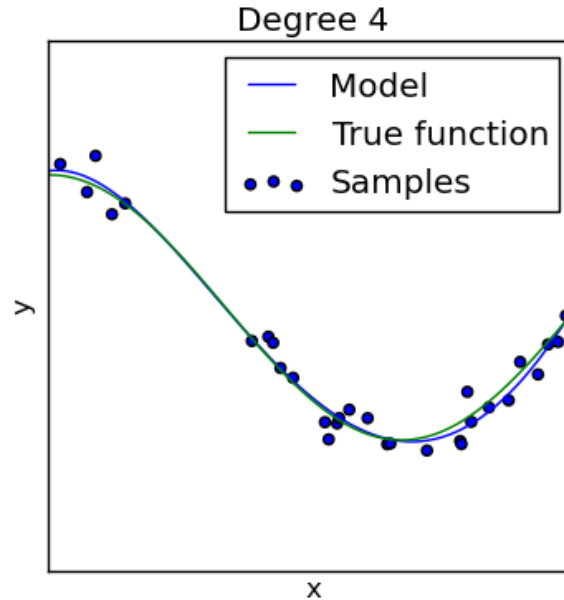


Diagnostics to detect under/overfitting

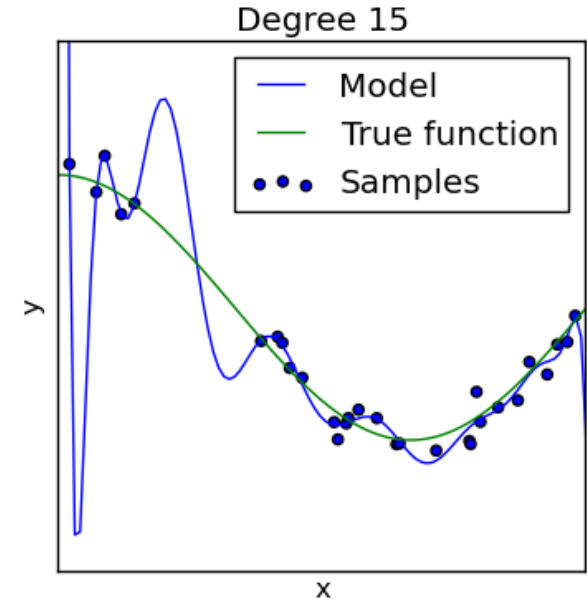
Underfitting



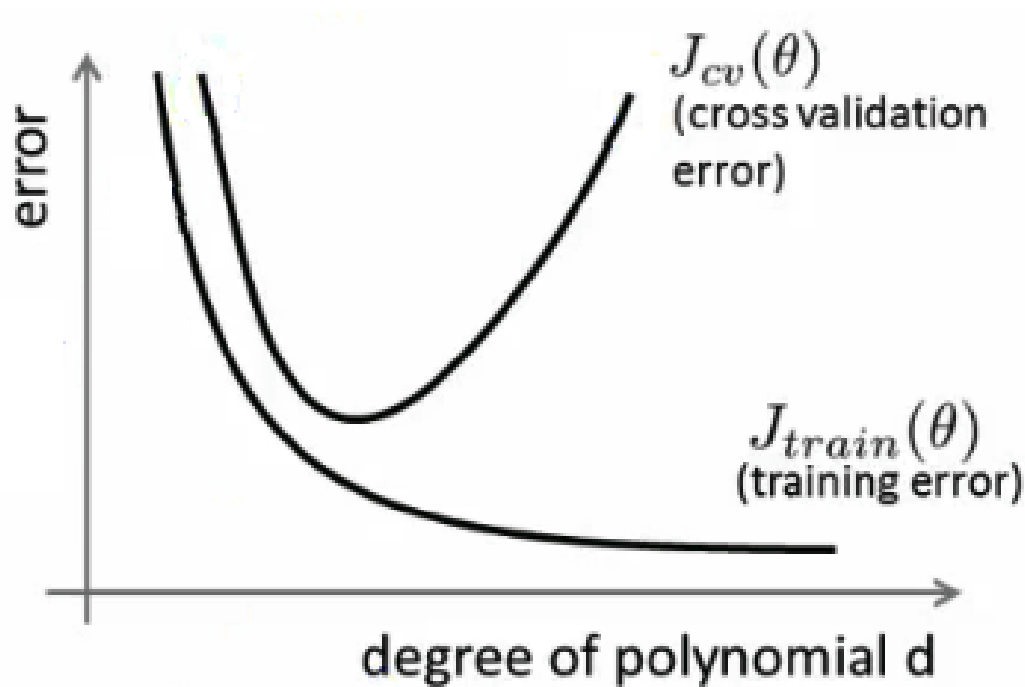
Just Right



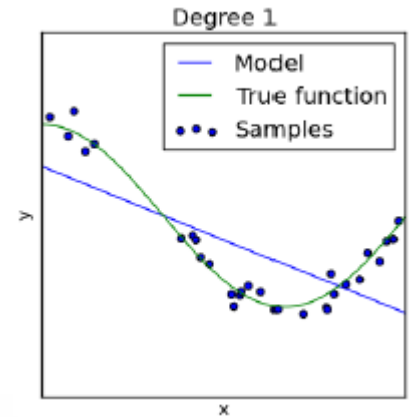
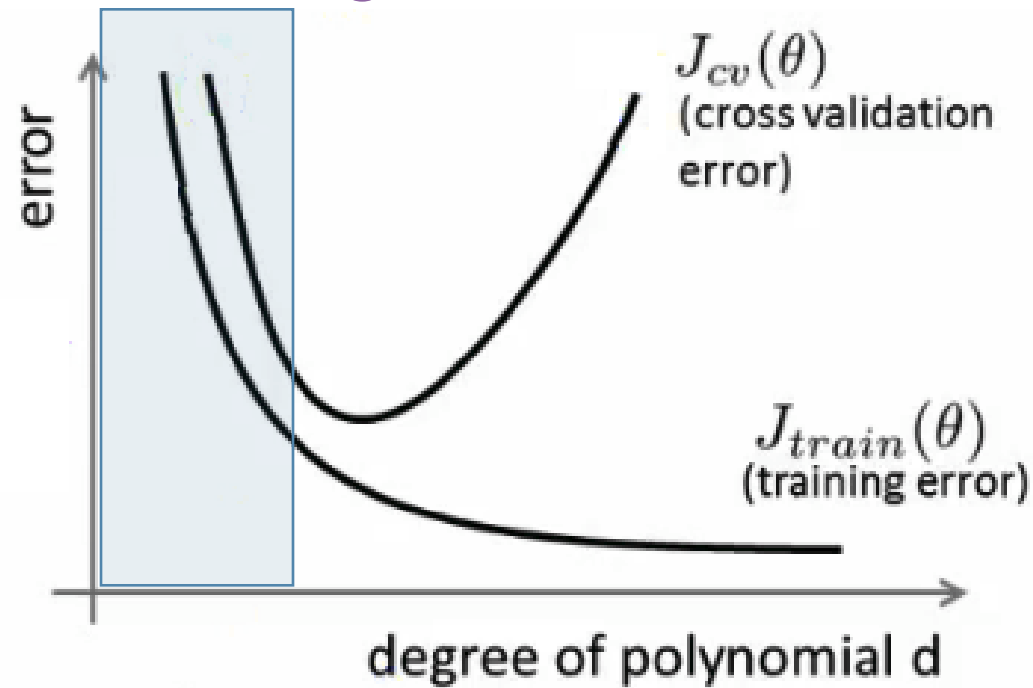
Overfitting



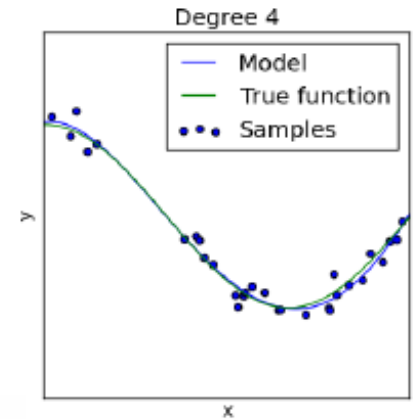
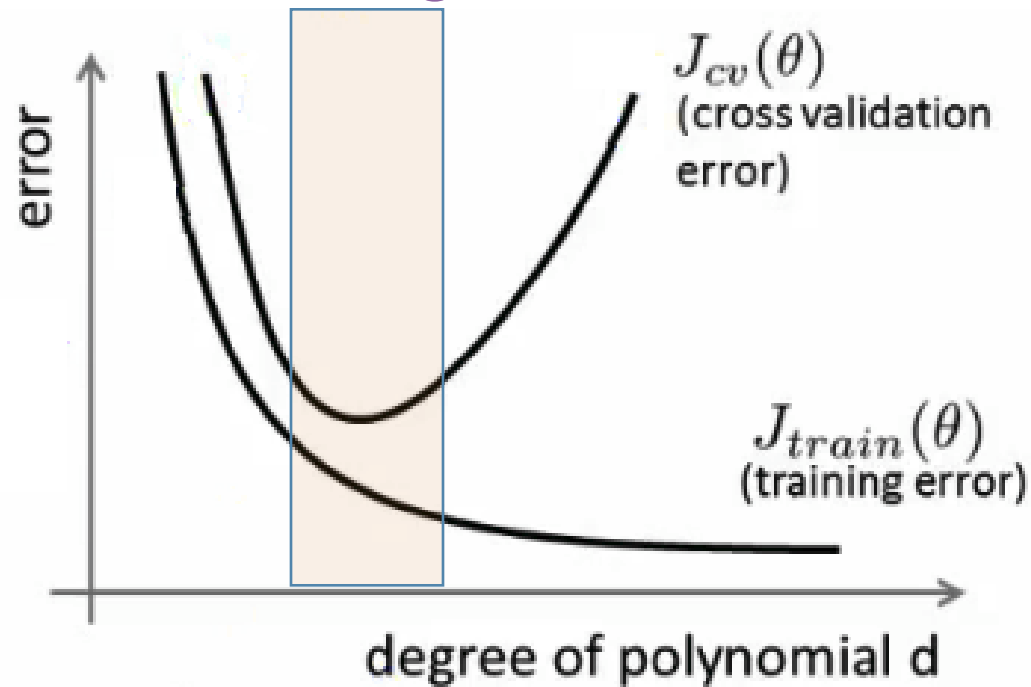
Model complexity vs error



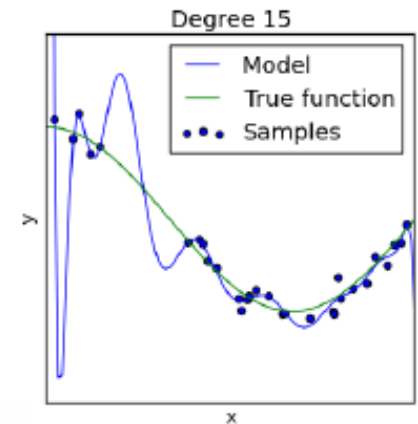
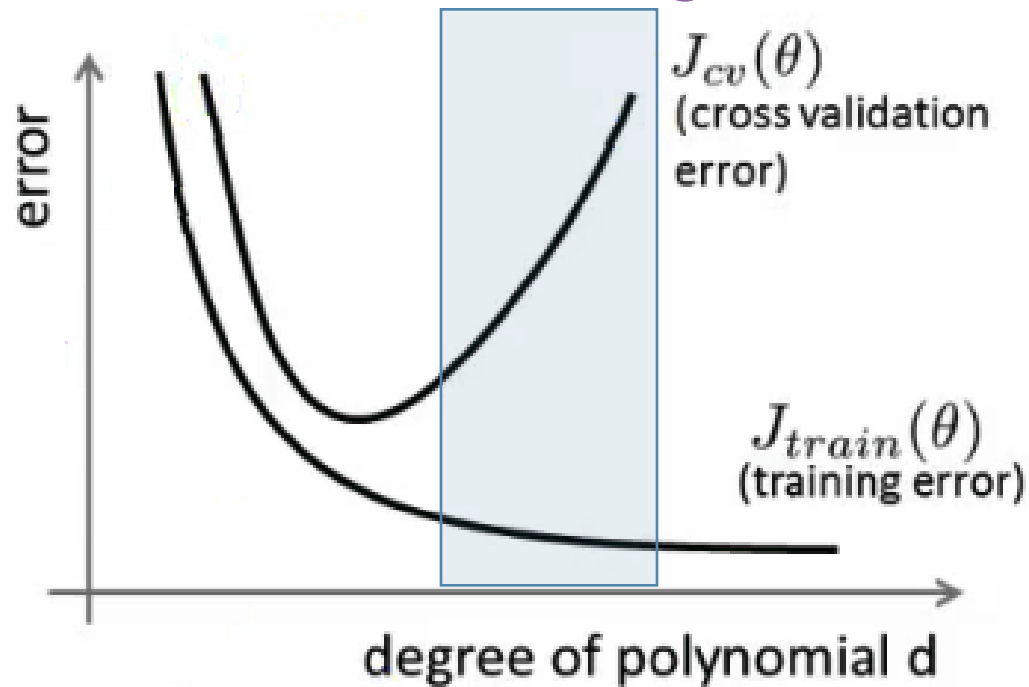
Underfitting



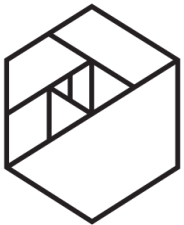
Just Right



Overfitting



Regularization



METIS

datascopes

OLS Regression Results

Dep. Variable:	DomesticTotalGross	R-squared:	0.286
Model:	OLS	Adj. R-squared:	0.278
Method:	Least Squares	F-statistic:	34.82
Date:	Sun, 14 Sep 2014	Prob (F-statistic):	6.80e-08
Time:	21:59:46	Log-Likelihood:	-1738.1
No. Observations:	89	AIC:	3480.
Df Residuals:	87	BIC:	3485.
Df Model:	1		

	coef	std err	t	P> t 	[95.0% Conf. Int.]
Budget	0.7846	0.133	5.901	0.000	0.520 1.049
Ones	4.44e+07	1.27e+07	3.504	0.001	1.92e+07 6.96e+07

Omnibus:	39.749	Durbin-Watson:	0.674
Prob(Omnibus):	0.000	Jarque-Bera (JB):	99.441
Skew:	1.587	Prob(JB):	2.55e-22
Kurtosis:	7.091	Cond. No.	1.54e+08

$$AIC = 2k - 2\ln(L)$$



parameters

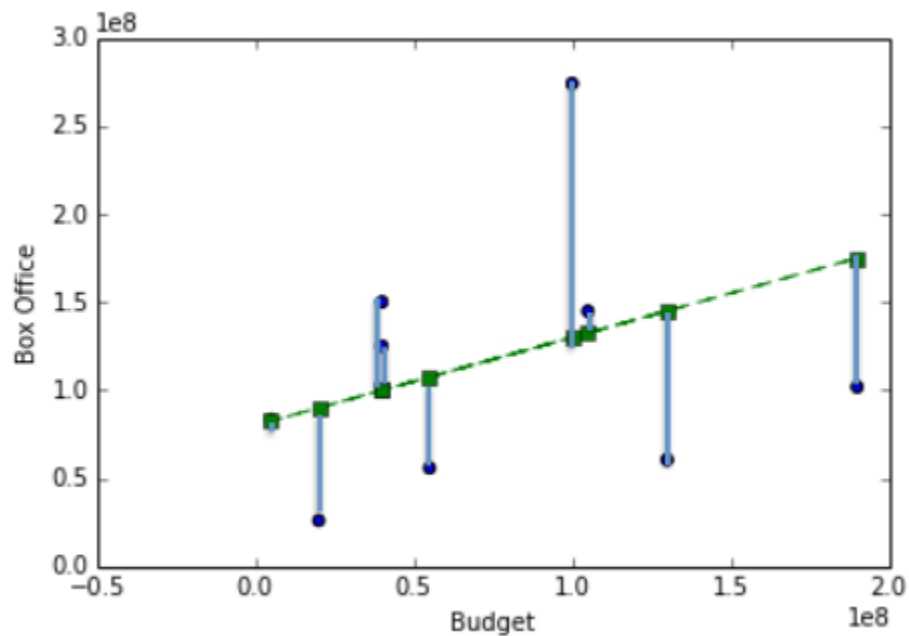


Log
likelihood

While awarding goodness of fit, penalize model complexity

While awarding goodness of fit, penalize model complexity

Why not do that while we are fitting?

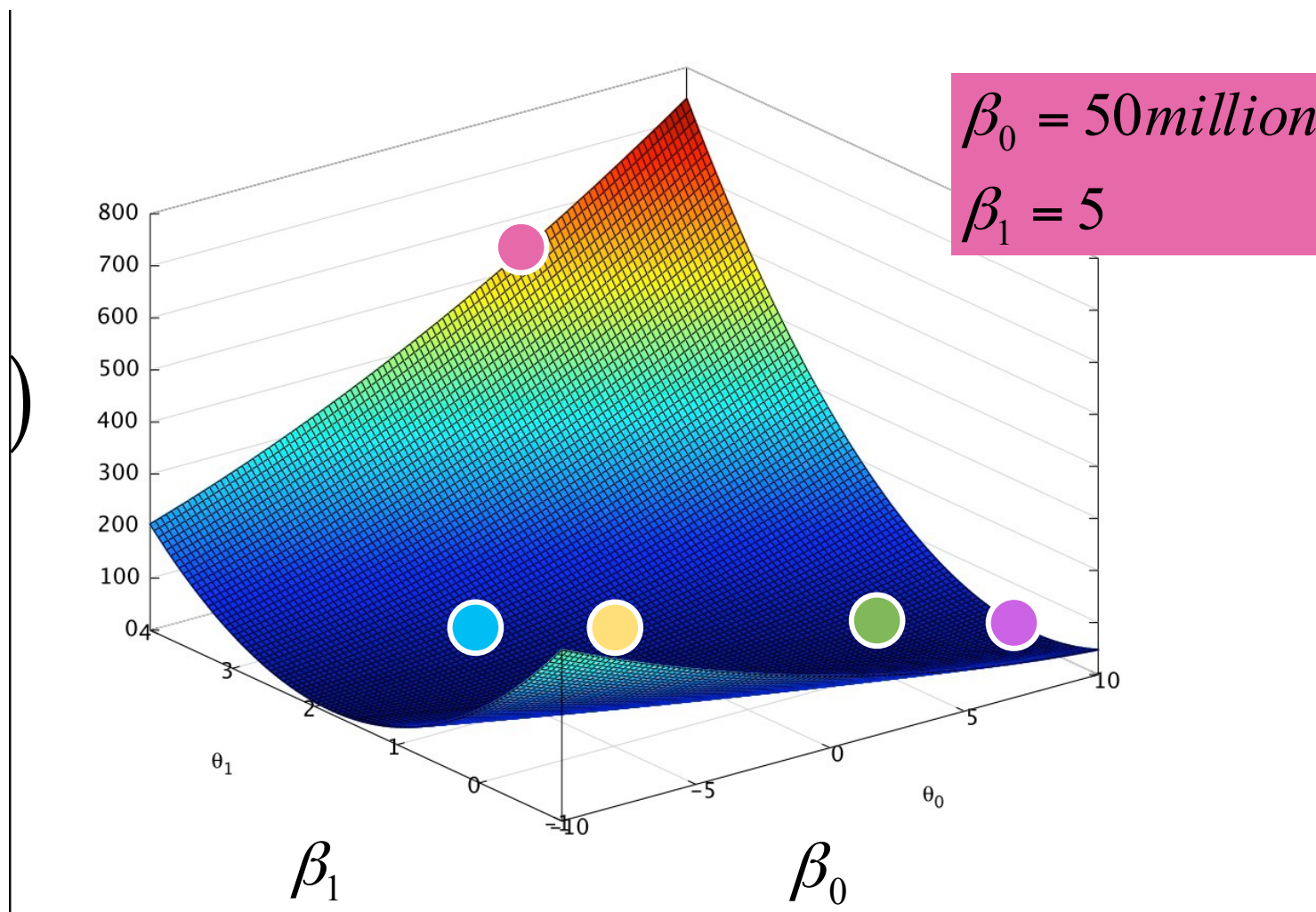


Cost function

Takes a model (specific parameter values), returns a score

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left((\beta_0 + \beta_1 x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2$$

$$J(\beta_0, \beta_1)$$



$\beta_0 = 80\text{million}$
 $\beta_1 = 0.5$

$\beta_0 = 0$
 $\beta_1 = 1.5$

$\beta_0 = 120\text{million}$
 $\beta_1 = 0.1$

$\beta_0 = 30\text{million}$
 $\beta_1 = 2$

Cost function

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left((\beta_0 + \beta_1 x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2$$



Lower for
better fits

Cost function

Add a penalty for the size of each parameter!

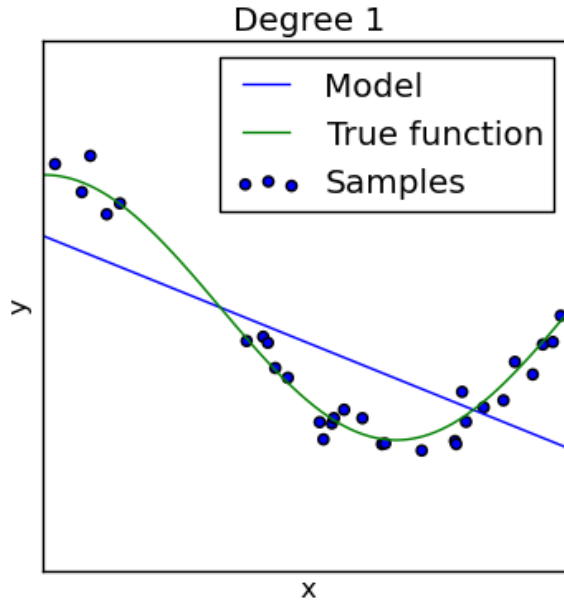
$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

Low: good fit
High: bad fit

Low: simple model
High: complex model

Diagnostics to detect under/overfitting

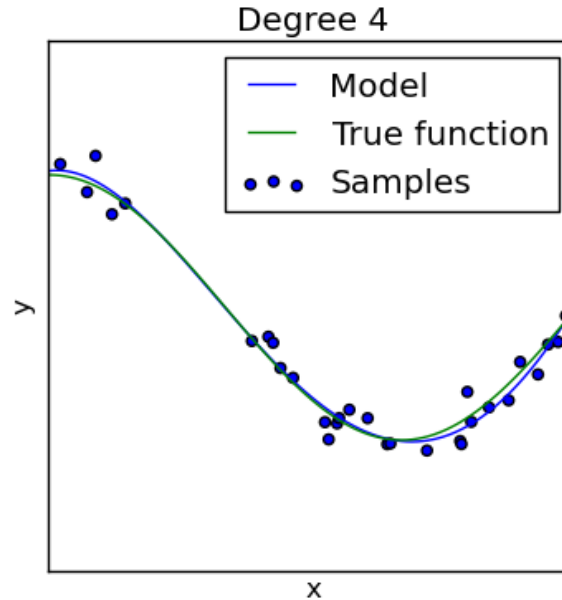
Underfitting



$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m (y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)})^2 + \lambda \sum_{j=1}^k \beta_j^2$$

J = V. High + Low

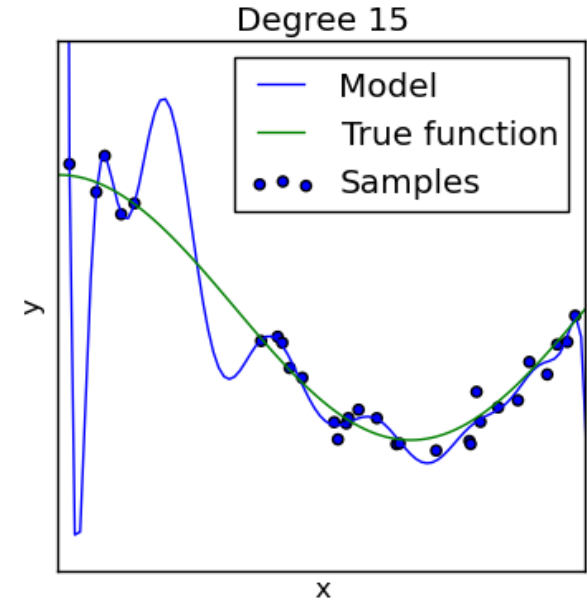
Just Right



$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m (y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)})^2 + \lambda \sum_{j=1}^k \beta_j^2$$

J = Low + Low

Overfitting



$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m (y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)})^2 + \lambda \sum_{j=1}^k \beta_j^2$$

J = Low + V. High

Ridge Regression

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

Underfitting

$$J = \text{V. High} + \text{Low}$$

Just Right

$$J = \text{Low} + \text{Low}$$

Overfitting

$$J = \text{Low} + \text{V. High}$$

$$\lambda = 1$$



$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

$$y_{\beta}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \varepsilon$$

Underfitting

$$J = \text{V. High} + \text{Low}$$

Just Right

$$J = \text{Low} + \text{Low}$$

Overfitting

$$J = \text{Low} + \text{V. High}$$

$$\lambda = 1$$



$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

$$\approx 0$$



$$\approx 0$$



$$y_{\beta}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \varepsilon$$

Underfitting

Just Right

Overfitting

$J = \text{V. High} + \text{Medium}$

$J = \text{Low} + \text{V High}$

$J = \text{Low} + \text{V V V High}$

VERY LARGE
underfit

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

≈ 0



≈ 0



≈ 0



≈ 0



$$y_{\beta}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \varepsilon$$

Underfitting

$$J = \text{V. High} + \text{Tiny}$$


Just Right

$$J = \text{Low} + \text{Tiny}$$

Overfitting

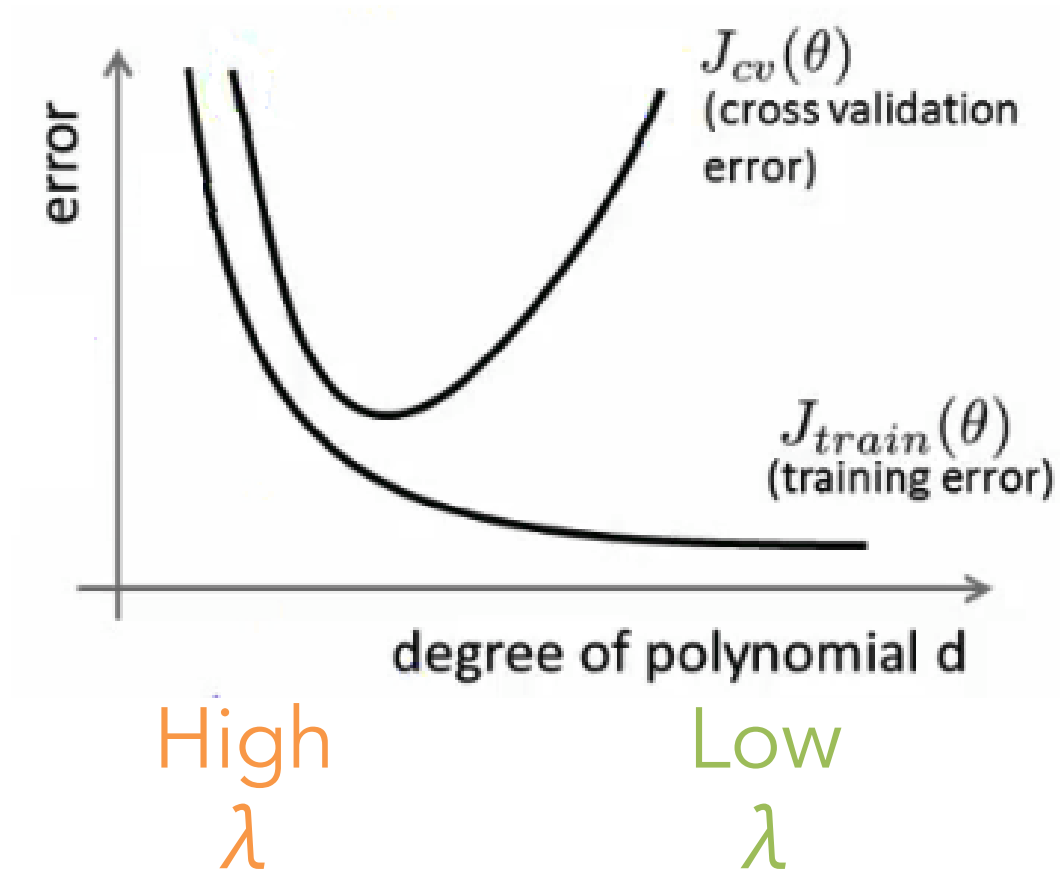
$$J = \text{Low} + \text{Tiny}$$

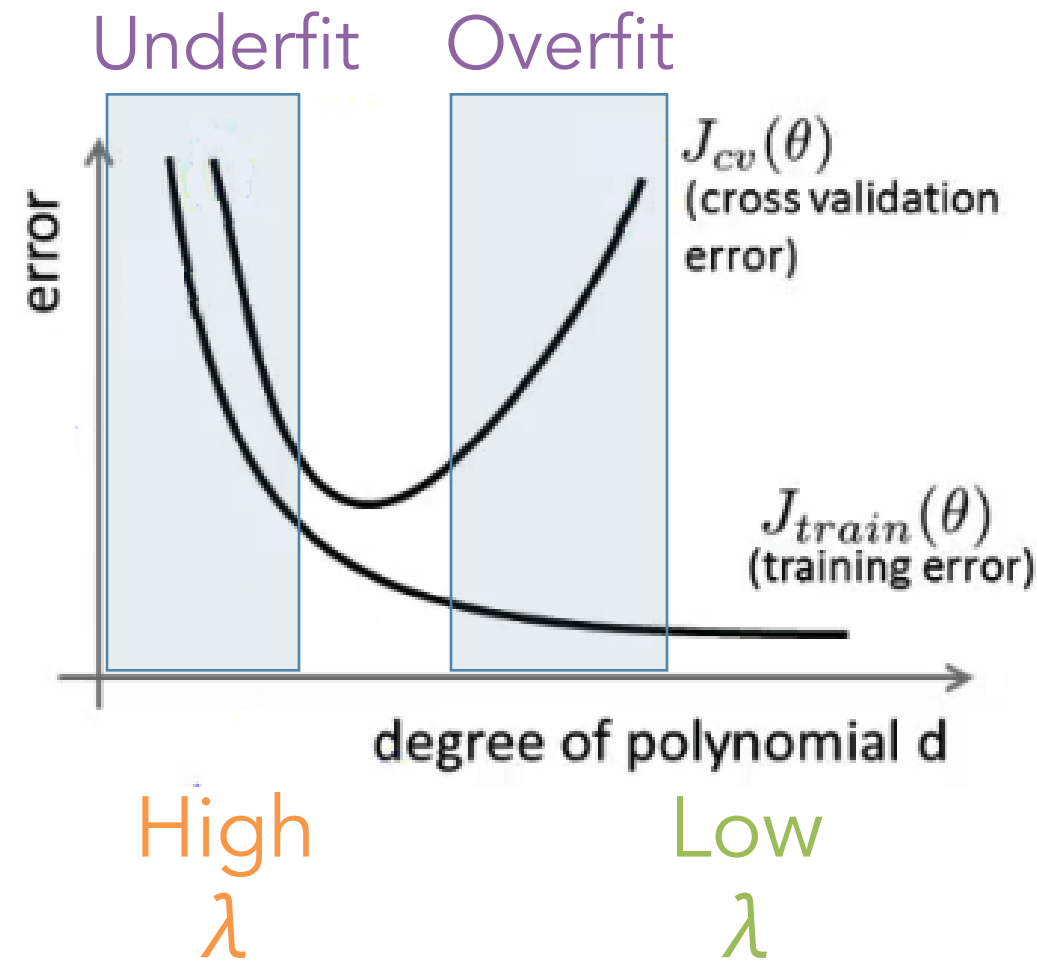
very small
possible
overfit


$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

$$y_{\beta}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \varepsilon$$

Error vs. regularization λ





Ridge Regularization (L2)

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

Ridge Regularization (L2)

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

Lasso Regularization (L1)

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k |\beta_j|$$

Ridge Regularization (L2)

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k \beta_j^2$$

Lasso Regularization (L1)

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda \sum_{j=1}^k |\beta_j|$$

Elastic Net (L1 + L2)

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left(y_{\beta}(x_{obs}^{(i)}) - y_{obs}^{(i)} \right)^2 + \lambda_1 \sum_{j=1}^k |\beta_j| + \lambda_2 \sum_{j=1}^k \beta_j^2$$

How can I use this?

We were doing:

```
from sklearn.linear_model import  
LinearRegression  
model = LinearRegression()  
model.fit(X,Y)
```

How can I use this?

We were doing:

```
from sklearn.linear_model import LinearRegression  
model = LinearRegression()  
model.fit(X,Y)
```

To use Ridge Regularization:

```
from sklearn.linear_model import Ridge  
model = Ridge(1.0)  
model.fit(X,Y)
```



λ (sklearn Calls It alpha)

How can I use this?

We were doing:

```
from sklearn.linear_model import LinearRegression  
model = LinearRegression()  
model.fit(X,Y)
```

To use Lasso:

```
from sklearn.linear_model import Lasso  
model = Lasso(1.0)  
model.fit(X,Y)
```



λ (sklearn Calls It
alpha)

How can I use this?

We were doing:

```
from sklearn.linear_model import LinearRegression  
model = LinearRegression()  
model.fit(X,Y)
```

To use Elastic Net:

```
from sklearn.linear_model import ElasticNet  
model = ElasticNet(1.0, l1_ratio = 0.5)  
model.fit(X,Y)
```



total weight for the full
penalty term



ratio of l1/l2 penalty



My model is not
awesome
enough.

What do I do?

Try these and check test error (and AIC,BIC,etc.) again:

Use a smaller set of features

Try adding polynomials

Check functional forms for each feature

Try including other features

Use more data (bigger training set)

Regularization

Try these and check test error (and AIC,BIC,etc.) again:

Use a smaller set of features

Try adding polynomials

Check functional forms for each feature

Try including other features

Use more data (bigger training set)

Regularization: Increase/decrease λ