

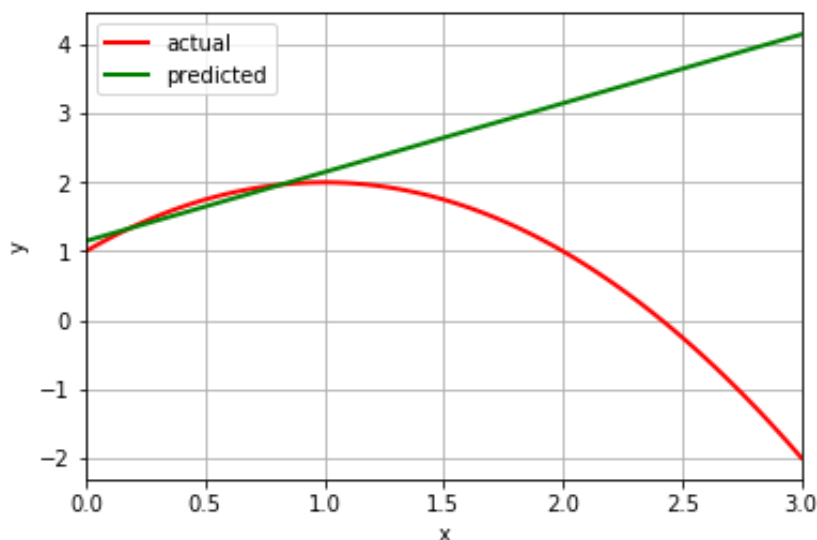
```
In [8]: 1 import numpy as np
        2 import matplotlib
        3 import matplotlib.pyplot as plt
        4 from sklearn import datasets, linear_model, preprocessing
        5 import numpy.polynomial.polynomial as poly
        6 %matplotlib inline
        7
        8
        9 beta = np.array([1, 2, -1])
       10 nsamp = 10
       11 xdat = np.linspace(0, 1, nsamp)
       12 ydat = poly.polyval(xdat, beta)
       13 d = 1
       14 beta_hat = poly.polyfit(xdat, ydat, d)

[ 1.14814815  1.          ]
```

```
In [14]: 1 xp = np.linspace(0, 3, 100)
        2 yp = poly.polyval(xp, beta)
        3 yp_hat = poly.polyval(xp, beta_hat)
```

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In [21]: 1 plt.xlim(0, 3)
        2 plt.plot(xp, yp, 'r-', linewidth=2)
        3 plt.plot(xp, yp_hat, 'g-', linewidth=2)
        4 plt.legend(['actual', 'predicted'], loc = 'upper left')
        5 plt.grid()
        6 plt.xlabel('x')
        7 plt.ylabel('y')
```

Out[21]: <matplotlib.text.Text at 0x1151fc400>



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In [22]: 1 bias = np.zeros(100)
          2 for i in range(100):
          3     bias[i] = (yp_hat[i] - yp[i])**2
          4 bias_order = np.argsort(-bias)
          5 print(xp[bias_order[0]])
```

3.0

```
In [ ]: 1
```