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In [1]: import numpy as np
import matplotlib.pyplot as plt

np.random.seed(1024) # ensure same noise for each run

# number of training points
n = 50

# sample n random points between 0 and 1
x = np.random.rand(n,1)

# set d = x^2 + .4 sin(1.5 pi x) + noise
d = x*x + 0.4*np.sin(1.5*np.pi*x) + 0.04*np.random.randn(n,1)

# plot result
plt.plot(x,d,'bo')
plt.xlabel('x')
plt.ylabel('d')
plt.title('Measured Data with Noise')
plt.show()
```

<Figure size 640x480 with 1 Axes>

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In [17]: sigma = 0.04 #defines Gaussian kernel width
p = 100 #number of points on x-axis

# Display examples of the kernels
x_test = np.linspace(0,1.00,p) # uniformly sample interval [0,1]
j_list = [5, 36, 46, 96] #list of indices for example kernels

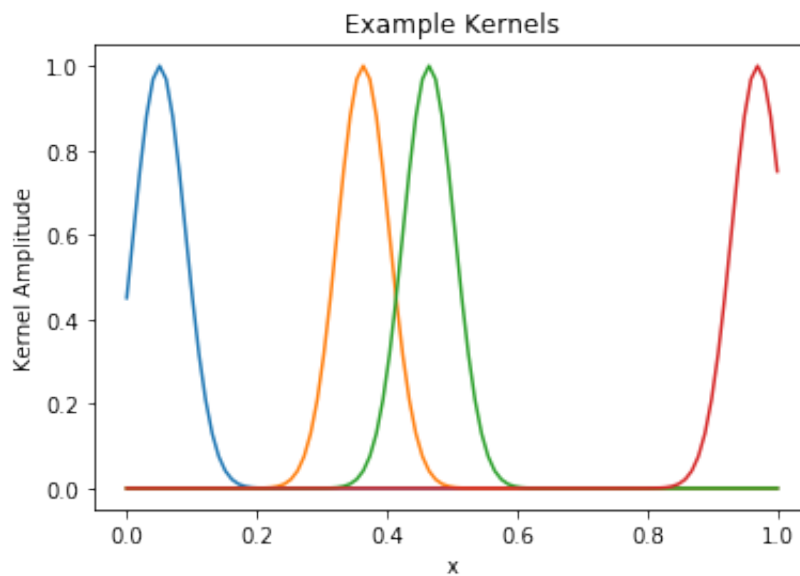
Kdisplay = np.zeros((p,len(j_list)),dtype=float)

for i in range(p):
    for j in range(len(j_list)):
        Kdisplay[i,j]= np.exp(-(x_test[i]-x_test[j_list[j]])**2/(2*sig

print('Sigma = ',sigma)
plt.plot(x_test, Kdisplay)
plt.title('Example Kernels')
plt.xlabel('x')
plt.ylabel('Kernel Amplitude')
plt.show()

```

Sigma = 0.04



```
In [18]: # Kernel fitting to data

lam = 0.01 #ridge regression parameter

distsq=np.zeros((n,n),dtype=float)

for i in range(0,n):
    for j in range(0,n):
        distsq[i,j]=(x[i]-x[j])**2

K = np.exp(-distsq/(2*sigma**2))

alpha = np.linalg.inv(K+lam*np.identity(n))@d
```

In [19]: *# Generate smooth curve corresponding to data fit*

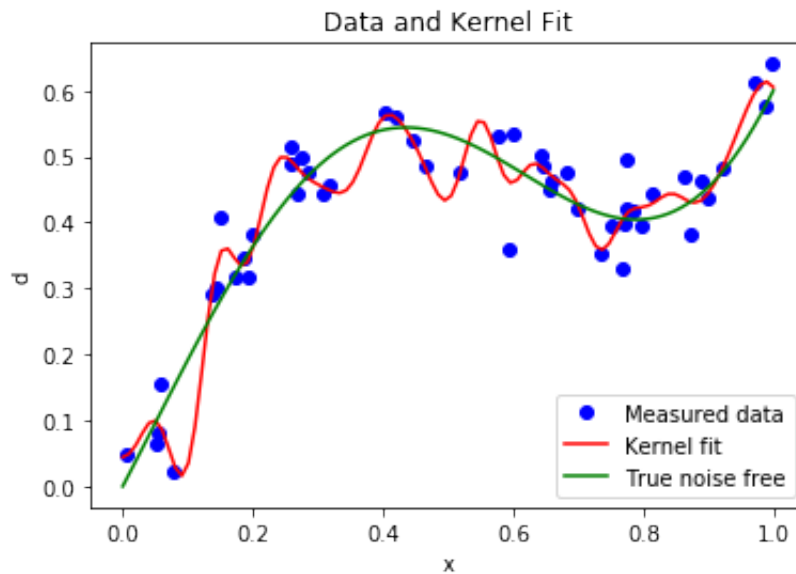
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distsq_xtest = np.zeros((p,n),dtype=float)
for i in range(0,p):
    for j in range(0,n):
        distsq_xtest[i,j] = (x_test[i]-x[j])**2

dtest = np.exp(-distsq_xtest/(2*sigma**2))@alpha

dtrue = x_test*x_test + 0.4*np.sin(1.5*np.pi*x_test) # noise free data

print('Sigma = ',sigma)
print('Lambda = ',lam)
plt.plot(x,d,'bo',label='Measured data')
plt.plot(x_test,dtest,'r',label='Kernel fit')
plt.plot(x_test,dtrue,'g',label='True noise free')
plt.title('Data and Kernel Fit')
plt.legend(loc='lower right')
plt.xlabel('x')
plt.ylabel('d')
plt.show()
```

Sigma = 0.04
Lambda = 0.01



1b)

ξ is around 0.3

ξ determines the actual position of a certain peak

σ determines the width between peaks

1d)

use cross validation to find the best σ and λ that generalizes well to new data.

In []: