

Lab 4 - Wyatt Madden & Dan Crowley

February 16, 2020

1 3.1

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In [1]: import scipy.io as scipy
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import numpy.linalg as lg
from eval_basis import *
from func_gauss import *
from func_hat import *
from gauss_basis import *
from hat_basis import *
```

```
In [2]: # Least-Squared Error FIT
# Find the linear combination of basis functions which best model the data.
#
# Inputs:
#
# x - Vector with observation locations in 1D. (indep. variable)
# t - Vector with observations in 1D. (dep. variable)
# params - Parameters for the basis functions to be used in func, e.g. as
# produced by gauss_basis.
# func - Function handle which evaluates a basis function with parameters
# given by the columns of params and at the specified locations. e.g.
# @gauss_basis, or @hat_basis.
# For example, the first basis function at x = 2 is func(2, params(:,1)).
# mu - Scalar representing the standard deviation of the prior Gaussian on
# the model parameters.
#
# Outputs:
# w - Coefficients used to generate a linear combination of the basis
# functions which is the maximum likelihood learned model.

def lsefit(x, t, params, func, mu):
    design_matrix = better_eval_basis(params = params,
                                      func = func,
```

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        xeval = x)

w_hat = lg.inv(np.dot(np.transpose(design_matrix), design_matrix) +
                 np.identity(design_matrix.shape[1])*(1/mu**2))
w = np.dot(w_hat, np.dot(np.transpose(design_matrix), t))
return w

In [3]: lab_4_dat = scipy.loadmat('/Users/wyattmadden/Documents/school/' +
                                   'MSU/2020/spring/m508/lab_info/lab_4/simple.mat',
                                   squeeze_me = True)

x = lab_4_dat['x']
t = lab_4_dat['t']

2 3.2

In [43]: hat_ten_10000 = lsefit(x = x,
                                t = t,
                                params = hat_basis(0, 2*np.pi, len(x)),
                                func = func_hat,
                                mu = 10**5)

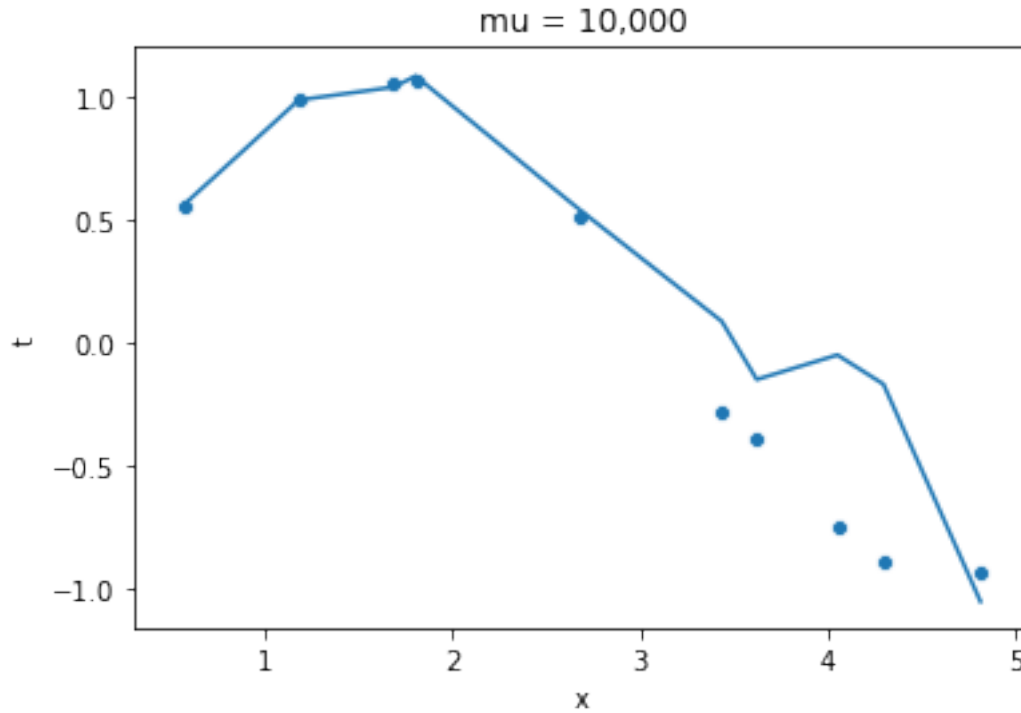
hat_df = {'hat_ten_10000': np.dot(np.transpose(hat_ten_10000),
                                     np.transpose(better_eval_basis(hat_basis(0, 2*np.pi,
                                     func_hat,
                                     x))),
                                'x': x,
                                't': t}

hat_df = pd.DataFrame(hat_df)

sns.lineplot(x = "x", y = "hat_ten_10000", data = hat_df)
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 10,000")

Out[43]: [Text(0.5,1,'mu = 10,000')]

```



The fit of the hat basis function with μ of 10,000 is not a good approximation of the data. It is too responsive to subsequent data points, especially on the lower end of the x space.

3 3.3

```
In [44]: hat_ten_10 = lsefit(x = x,
                             t = t,
                             params = hat_basis(0, 2*np.pi, len(x)),
                             func = func_hat,
                             mu = 10)

hat_ten_1 = lsefit(x = x,
                  t = t,
                  params = hat_basis(0, 2*np.pi, len(x)),
                  func = func_hat,
                  mu = 1)

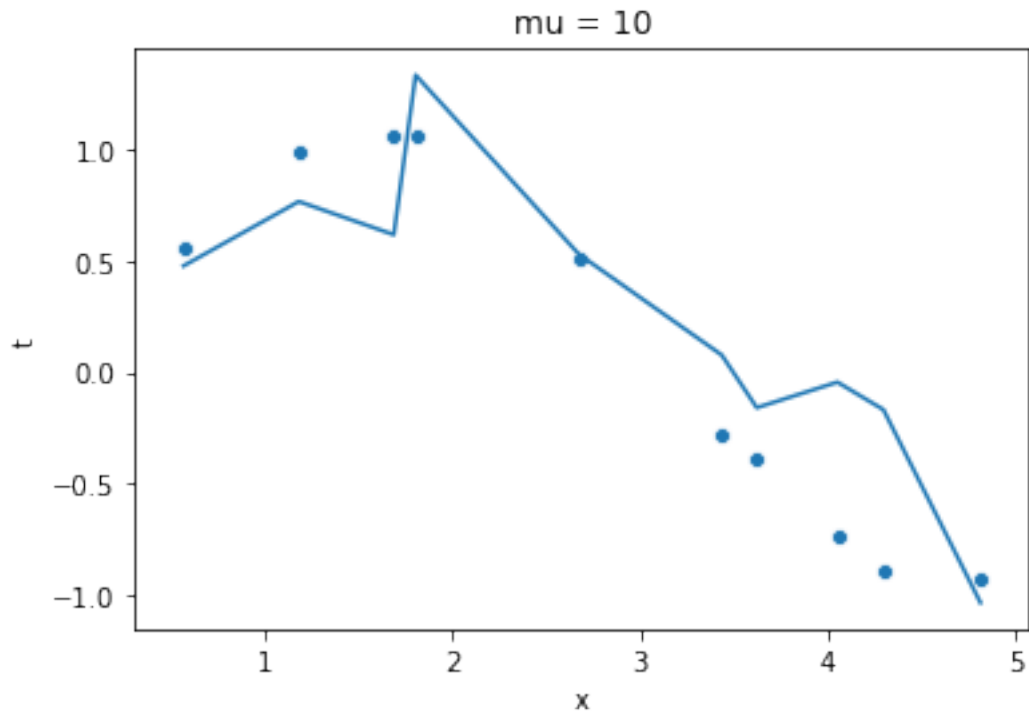
hat_df = {'hat_ten_10': np.dot(np.transpose(hat_ten_10),
                                np.transpose(better_eval_basis(hat_basis(0, 2*np.pi,
                                                                    func_hat,
                                                                    x)))),
          'hat_ten_1': np.dot(np.transpose(hat_ten_1),
                                np.transpose(better_eval_basis(hat_basis(0, 2*np.pi,
                                                                    func_hat,
                                                                    x))))),
```

```
'x': x,  
't': t}
```

```
hat_df = pd.DataFrame(hat_df)
```

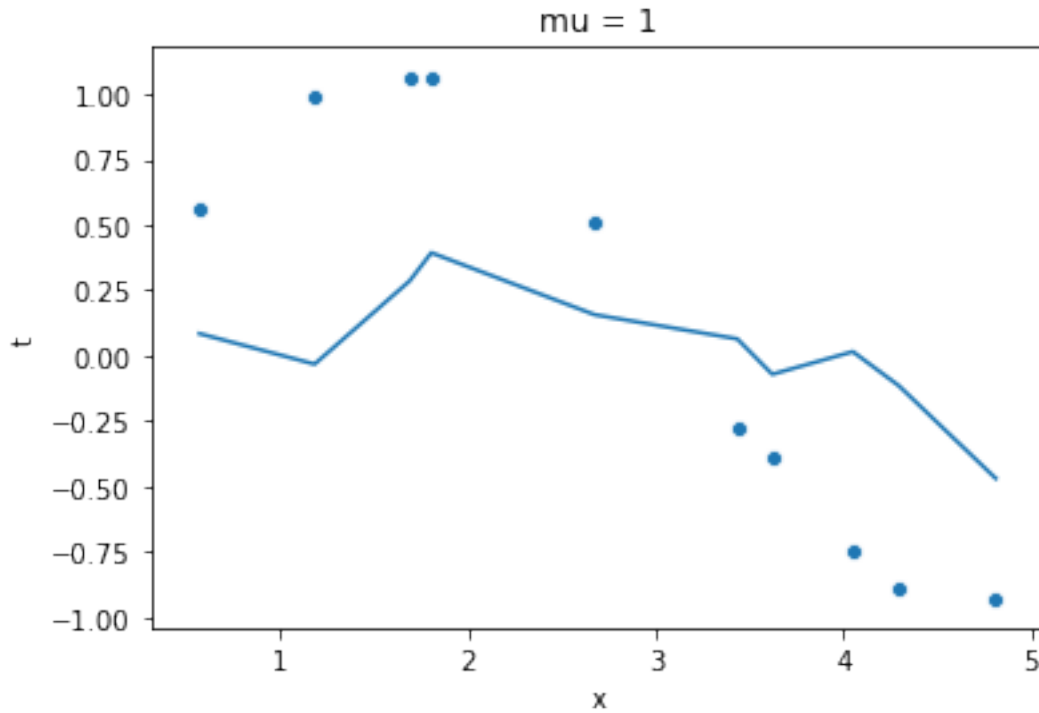
```
sns.lineplot(x = "x", y = "hat_ten_10", data = hat_df)  
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 10")
```

```
Out[44]: [Text(0.5,1,'mu = 10')]
```



```
In [45]: sns.lineplot(x = "x", y = "hat_ten_1", data = hat_df)  
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 1")
```

```
Out[45]: [Text(0.5,1,'mu = 1')]
```



4 3.4

5 3.5

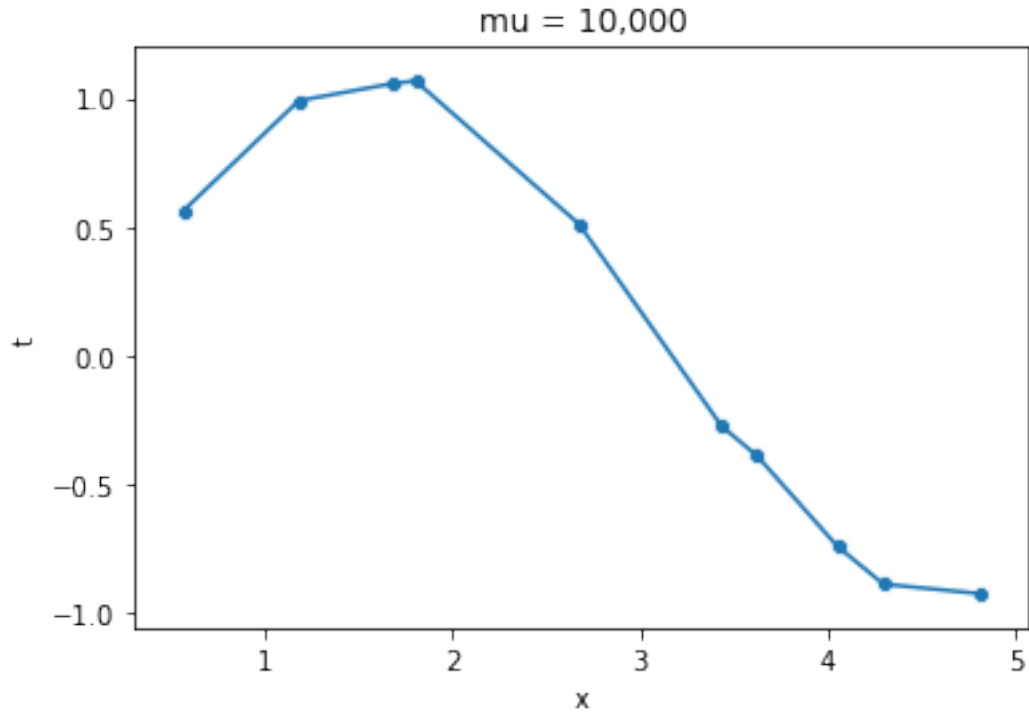
```
In [48]: gauss_ten_10000 = lsefit(x = x,
                                   t = t,
                                   params = gauss_basis(0, 2*np.pi, len(x)),
                                   func = func_gauss,
                                   mu = 10**5)

gauss_df = {'gauss_ten_10000': np.dot(np.transpose(gauss_ten_10000),
                                       np.transpose(better_eval_basis(gauss_basis(0, 2*np.pi,
                                       func_gauss,
                                       x))),
                                   'x': x,
                                   't': t}

gauss_df = pd.DataFrame(gauss_df)

sns.lineplot(x = "x", y = "gauss_ten_10000", data = gauss_df)
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 10,000")

Out[48]: [Text(0.5,1,'mu = 10,000')]
```



6 3.6

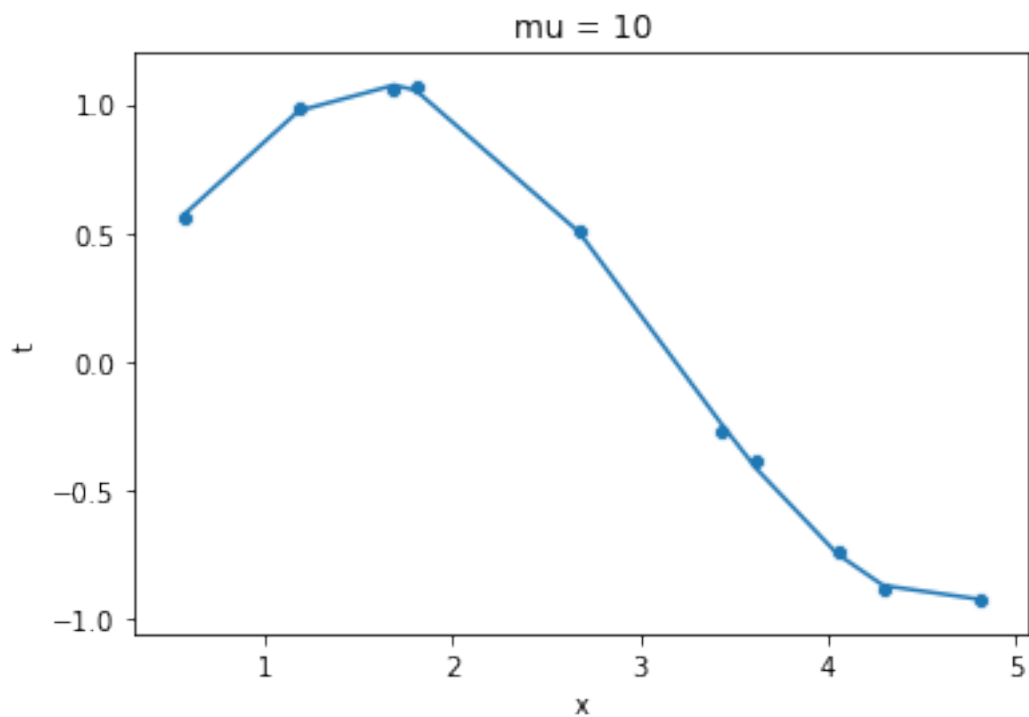
```
In [49]: gauss_ten_10 = lsefit(x = x,
                                t = t,
                                params = gauss_basis(0, 2*np.pi, len(x)),
                                func = func_gauss,
                                mu = 10)

gauss_df = {'gauss_ten_10': np.dot(np.transpose(gauss_ten_10),
                                    np.transpose(better_eval_basis(gauss_basis(0, 2*np.pi,
                                                                    func_gauss,
                                                                    x)))),
            'x': x,
            't': t}

gauss_df = pd.DataFrame(gauss_df)

sns.lineplot(x = "x", y = "gauss_ten_10", data = gauss_df)
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 10")

Out[49]: [Text(0.5,1,'mu = 10')]
```



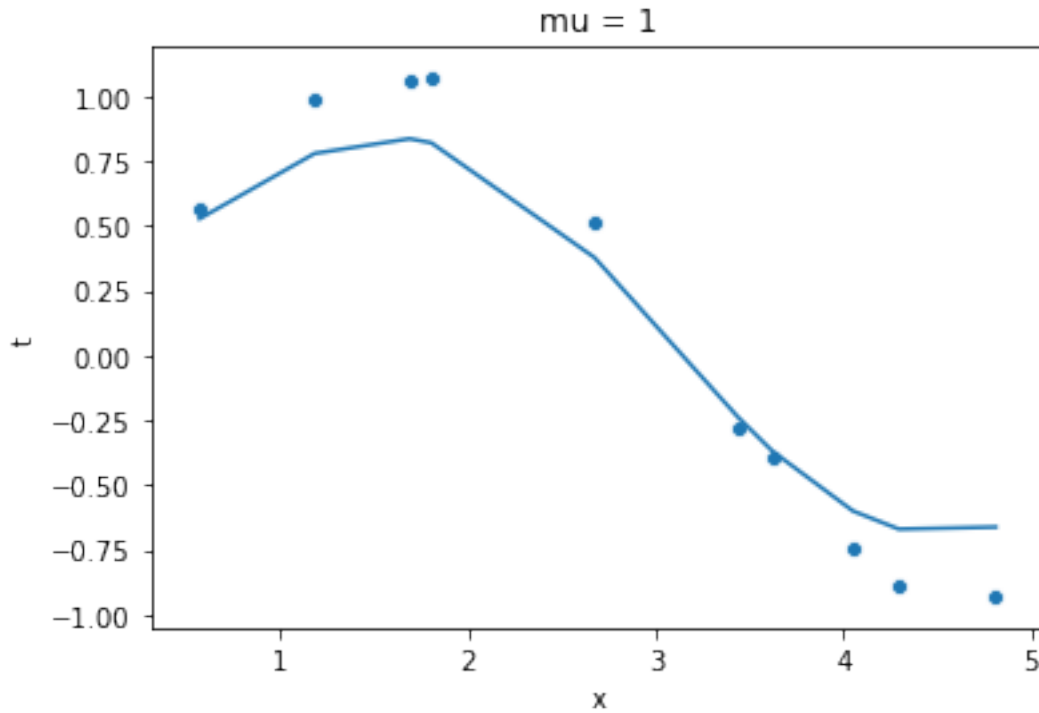
```
In [50]: gauss_ten_1 = lsefit(x = x,
                             t = t,
                             params = gauss_basis(0, 2*np.pi, len(x)),
                             func = func_gauss,
                             mu = 1)

gauss_df = {'gauss_ten_1': np.dot(np.transpose(gauss_ten_1),
                                   np.transpose(better_eval_basis(gauss_basis(0, 2*np.pi,
                                                                       func_gauss,
                                                                       x)))),
            'x': x,
            't': t}

gauss_df = pd.DataFrame(gauss_df)

sns.lineplot(x = "x", y = "gauss_ten_1", data = gauss_df)
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 1")

Out[50]: [Text(0.5,1,'mu = 1')]
```



7 3.7

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10 3.10

```
In [51]: gauss_ten_1 = lsefit(x = x,
                             t = t,
                             params = gauss_basis(0, 2*np.pi, 100),
                             func = func_gauss,
                             mu = 1)

gauss_df = {'gauss_ten_1': np.dot(np.transpose(gauss_ten_1),
                                   np.transpose(better_eval_basis(gauss_basis(0, 2*np.pi, 100),
                                                                     func_gauss, x))),
            'x': x,
            't': t}

gauss_df = pd.DataFrame(gauss_df)
```



```
sns.lineplot(x = "x", y = "gauss_ten_1", data = gauss_df)
sns.scatterplot(x = "x", y = "t", data = hat_df).set(title = "mu = 1")
```

Out[51]: [Text(0.5,1,'mu = 1')]

