

1. Curve fitting as an example of regression, using PMTK.
  - 1) Download and install PMTK from the following address:  
<https://github.com/probml/pmtk3>  
Follow its Readme to download and setup.
  - 2) Run the demo script `/demos/linregPolyVsDegree.m`
    - (a) Attach Figure 1 in your answer.
    - (b) Describe how curve shape and MSE change as degree of polynomial increases, and explain why.
  - 3) Run the demo script `/demos/linregPolyVsRegDemo.m`
    - (a) Attach Figure 1 in your answer.
    - (b) What is degree of polynomial? Is it changing during the demo?
    - (c) Which variable controls the effect of regularization? Describe how curve shape and MSE change as regularizer increases and why. **Hint:** regularization is introduced in Murphy 7.5.1 (part of Reading Assignment 1), and in Lecture 3.

2. Implement a simple curve fitting problem.

- 1) [\[code\]](#) Generate a set of data points with the following lines of Matlab code:

```
% a simple curve fitting script
x = [0:0.5:10]';
y = 0.03 * x.^3 - 0.4 * x.^2 + x + 0.2 + normrnd(0, 0.3, size(x));
```

We can see that  $y$  is a set of points drawn from a polynomial corrupted by Gaussian noise. Plot  $y$  as discrete points.

- 2) [\[math\]](#) Assume that we know the curve to be fit,  $f(x)$ , is a 3rd order polynomial. Write down the mean-squared error objective function (criterion function, or function to be minimized) for curve fitting. This is the function that enables the algorithm to learn from the data points.
- 3) [\[math\]](#) Represent the objective function in matrix form.
- 4) [\[code\]](#) Solve for the curve parameters using pseudo-inverse. Plot the resulting polynomial (as a continuous curve) along with  $y$  (as discrete points).
- 5) [\[code\]](#) what does your resulting fitted curve give for  $x=3.25$ ?  $x=9.75$ ?
- 6) [\[math\]](#) How do these values compare with the original expression for  $y$ , except without corrupting noise, at those points?

Attach your Matlab or Python code in your answers.