## Assignment 7

Corrections: [none yet]

1. We have points $x_{k}$ and values $f_{k}$. We want a polynomial of degree $d$ so that $p\left(x_{k}\right)=f_{k}$ for $k=0, \ldots, d$. Write a Python module that constructs $p(x)=p_{0}+{ }_{\_} 1 x+\cdots+p_{d} x^{d}$ by solving a linear system of equations for the coefficients $p_{k}$. Use $d+1$ points uniformly spaced in the interval $[-1,1]$. Create a Python module pFit.py that finds the coefficients, given the $x_{k}$ and $f_{k}$ by solving a linear system with the Vander Monde matrix. It also should return the condition number of the Vander Monde matrix. You can evaluate the condition number using a Nympy routine or from the singular values. Use the appropriate Numpy linear algebra routine to do the solving. Write another Python module pEval.py that takes the coefficients $p_{k}$ and a point $x$ and evaluates $p(x)$. Do this with Horner's rule.
(a) Write a module pTest.py that does several tests to see that the other two routines work correctly. It should test both routines.
(b) Write a module that plots the condition number of the VanderMonde matrix as a function of $d$. Make a plot (log-log or semi-log or whatever) that makes clear the quantitative behavior of the condition number as $d$ grows. Does it grow with $d$ ? Does it grow polynomially or exponentially? Think about what kind of plot would answer these questions and make it.
(c) Plot the polynomial fits of

$$
F(x)=e^{-\frac{1}{2} x^{2}}
$$

Plot the polynomial of degree $d$ that interpolates $F$ at $d+1$ uniformly spaced points in the interval $[-2,2]$. (Warning: this looks like the Runge example but it isn't.) For plotting, you need a large number of uniformly spaced points (maybe a few hundred?), even if $d$ is much smaller than that. Comment on the error as a function of $d$. Comment on the error inside and outside of the interval $[-1,1]$ that was used for fitting. What effect does the conditioning of the Vander Monde matrix have? Please answer quantitatively.
2. Write Python modules sFit.py, sEval.py and sTest.py that repeat much of Exercise 1, but with cubic B-splines. The fitting routine takes $d+1$ knot points $x_{0}<x_{1} \leq x_{d}$ and constructs a piecewise cubic polynomial $s(x)$ so that for $x_{k} \leq x \leq x_{k+1}$ the spline is a cubic polynomial

$$
s(x)=q_{k}(x)=a_{k}+b_{k}\left(x-x_{k}\right)+c_{k}\left(x-x_{k}\right)^{2}+d_{k}\left(x-x_{k}\right)^{3} \text { for } x_{k} \leq x \leq x_{k+1}
$$

We use $\left(x-x_{k}\right)$ instead of $x$ to make the condition number better. The coefficients should be found from the following conditions

$$
\begin{aligned}
& s\left(x_{k}\right)=f_{k} \text { for } k=0, \ldots, d \\
& s^{\prime} \text { and } s^{\prime \prime} \text { are continuous } \\
& s^{\prime \prime}\left(x_{0}\right)=0 \\
& s^{\prime \prime}\left(x_{d}\right)=0
\end{aligned}
$$

Show that this is $4 d$ linear equations for the $4 d$ unknown coefficients. Repeat the steps from Exercise 1 to create spline interpolation software.
(a) The module sText. py should use finite differences on the computed $s(x)$ to verify that the fourth derivative is zero (for a cubic) except at knot points $x_{k}$. It should verify that $s(x)$ and $s^{\prime}(x)$ and $s^{\prime \prime}(x)$ is continuous at knot points (finite differences). It should verify that the interpolation conditions are satisfied.
(b) Plot the condition number of the $4 d \times 4 d$ matrix as a function of $d$. Contrast this condition number to the condition number of the matrix used for high order polynomial interpolation.
(c) Test the accuracy on $e^{-\frac{1}{2} x^{2}}$ on $[-1,1]$. Make a plot with a small but not too small number of knots where you can see the error. Make a plot of the max error as a function of $d$. If you have time, try to find the order of accuracy.
3. Radial basis function interpolation finds a function

$$
r(x)=\sum_{k=0}^{d} w_{k} \phi\left(x-x_{k}\right)
$$

The function $\phi(x)$ is the radial basis function. Take it to be the "quadratic exponential"

$$
\phi(x)=e^{-\frac{x^{2}}{2 L^{2}}}
$$

In more than one dimension, this would be $e^{-\frac{|x|^{2}}{2 L^{2}}}$, which is a function only of $r=|x|$ and explains the term "radial" basis function. Radial basis function interpolation involves a "length scale" parameter $L$. Write a set of Python modules rFit.py, and rEval.py, and rTest.py. The fitting routine should choose weights $w_{k}$ to satisfy interpolation conditions $r\left(x_{k}\right)=f_{k}$.
(a) The tester rTest.py should check that the interpolation conditions are satisfied.
(b) Explore the condition number as a function of $d$ and $L$ for uniformly spaced points.
(c) Do radial basis function interpolation on $e^{-\frac{1}{2} x^{2}}$ with $d+1$ points and length scale $L$ to see which combinations give good results.

