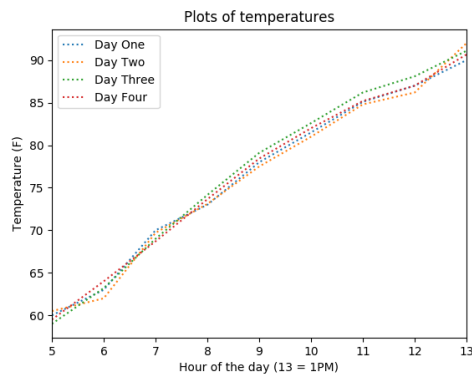


# Project 3 Report

## CMSC 409 - Artificial Intelligence

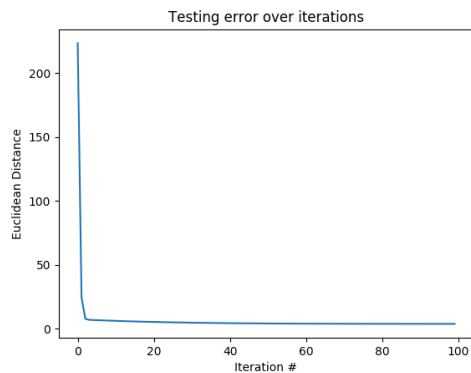
Steven Hernandez

1. There would be two input and one output for our unit. Inputs would be the hour and a bias input while output would be the estimated temperature at that hour of the day. In fact, because we have weights for  $x$  (hour of the day) and a bias, we can create the formula  $net = ax+b$  which means our unit can simply return  $net * 1$  or the identity.
2. The activation function would be some linear function. Our unit would not have a threshold however. Whatever the outcome from the linear activation function is would be the exact result from the learning unit. If we look at the graph of temperatures for our training (and testing) data, we can see that the values are basically just a linear function.

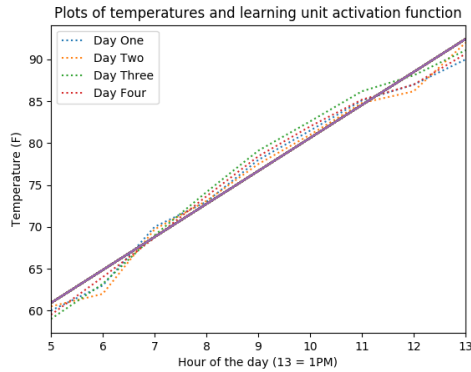


3. Outcome of training with days 1-3:

Euclidean distance comes down from 223.433536 to 3.789658



resulting in an activation as so:



4.

input	expected output	actual output	Euclidean distance
5	59.5	60.9108715013	1.41087150129
6	64	64.8531267022	0.8531267022
7	68.7	68.7953819031	0.0953819031114
8	73.65	72.737637104	-0.912362895977
9	78.43	76.6798923049	-1.75010769507
10	82	80.6221475058	-1.37785249415
11	85.2	84.5644027068	-0.635597293242
12	87	88.5066579077	1.50665790767
13	90.67	92.4489131086	1.77891310858

5. Learning rate was 0.0005 to keep the learning from going to quickly, while we went through 100 iterations.

Notice from the graph above on Euclidean distances, we reach our peak around the 20th iteration mark

6. As such, after the 20th iteration, we reach a plateau of improvement with our current system.

7. Using a more complex network with greater than one unit would allow for more complex output which would ultimately help us with this problem.

Currently, we are stuck with a linear output because the single unit can only learn as such.