Introduction to Neural Networks U. Minn. Psy 5038 Spring, 1999 Daniel Kersten

Problem Set 1

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#### **Exercise 1**

1A. Let w be a weight vector representing a template pattern. Let  $\{x\}$  be a collection of pattern vectors all of unit length. Show theoretically that the cross-correlator gives maximum response to the pattern which matches the form of the template pattern. Recall that the response is determined by the dot-product of the input vector with the weight vector. row that matches the input vector.

**1B.** Now write a *Mathematica* program to demonstrate this property of cross-correlators. Use the **Table** function to fill a 32x32 matrix **R** with random numbers. Use the built-in function **Random**[]. Then define a function **normalize**[**x**] that takes as input a vector **x**, and returns a normalized version of **x**. Use the **Table** function again to turn **R** into a matrix **R2** whose rows are normalized to unit length. Calculate the matrix product of **R2** with the 8th row of **R2**. Use **ListPlot** to show that the maximum of the product occurs at element 8. Make several more plots using other rows of **R2**, and show the maximum always occurs at the row that matches the input vector.

## **Exercise 2**

Use a set of rules to define a semi-linear "squashing" function, limit[x], which is:

```
-1 for x < -1;
x for 1 >=x >= -1;
1 for x> 1.
```

Plot **limit**[**x**] from x = -2 to 2.

### **Exercise 3**

Using *Mathematica*'s ability to find derivatives of functions, define a function **dsquash[]** to be equal to the derivative of the logistic function:

```
squash[r_] := 1/(1 + Exp[-r]);
```

Plot dsquash from r = -2 to 2.

(Hint: you may wish to use the rule for replacing a variable with a value in an expression. This would enable you to define the derivative function all on one line. Otherwise, you can compute the derivative, copy it, and then turn that copied cell into an **input cell** type. Use **Cell menu>Convert To**). Later on, we will need to use the derivative of the non-linear squashing function in our neural networks. For this reason, it is useful to have a squashing function that has a closed form solution for the derivative.

#### **Exercise 4**

There are neurons in the primary visual cortex of mammals called "simple cells". One model for these cells is a linear crosscorelator followed by a thresholding non-linearity (e.g. the half-wave rectification of a diode). The receptive field weights of this cross-correlator typically show a "center-surround" organization. In one dimension, a much reduced model weight vector could look like this:

$$w = \{-2, -1, 6, -1, -2\}$$

Define a threshold function **thresh**[s] that is zero for s less than zero, and equal to s for values of s greater than or equal to 0.

Use the above weight vector **w**, and your **thresh**[] function to model the response of a simple cell. What is the response of your cell to an input **x**:

a) **x** = {-1,-.5,3,-.5,-1}

or

b) **x** = {2, 1, 0, 1, 2} ?

#### Exercise 5 (Requires material in Lecture 4)

Express the vector:

$$h = \{1, 2, 3, 4, 5, 6, 7, 8\};$$

as a linear sum of normalized Walsh vectors (feel free to copy and paste code from Lecture 4). Plot the "spectrum" of **h**. In particular use **ListPlot** to show the spectrum, which consists of the eight values of the projections of **h** onto the 8 Walsh functions. Verify your answer by reconstructing **h** from the projections.

# Optional Exercises

## **Exercise 1**

Write a Mathematica function that computes the McCulloch-Pitts Inclusive OR and AND functions.

# Exercise 2

Write a Mathematica function that includes the noise term in the generic connectionist model.