QUIZ 1

1. Which is the right output? (B)

? X is 55 + (6 / 2).

A.
$$X = 55 + (6/2)$$
.

- B. X = 58
- C. Error

2. Which is the right output? (A)

data(fact). data(rule).data(list).
cut_test_a(X) :-data(X).
cut_test_a('last clause').
cut_test_b(X) : data(X),
!.
cut_test_b('last clause').

? -cut_test_b(X), write(X), nl, fail.

- A, fact
- B fact rule | last clause
- C. fact rule list last clause
- D. Error

3. Which is the right output? (A)

?-a(b,c,d) = a(X,X,d).

?-a(c,X,X) = a(Y,Y,b).

- A. no no
- B. no yes
- C. yes no
- D. yes yes

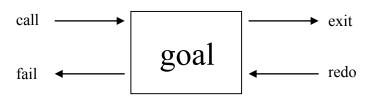
4.Terms / Data types in Prolog can be? (ABCD)

A. Integer

B. Atom

- C. Variable
- D. Structure

1. Please write the correct ports in Prolog goal?



6. Predict the results of these unification queries.

?-a(b,c) = a(X,Y).

X = b,

Y = c

?- a(X,c(d,X)) = a(2,c(d,Y)).

X = Y, Y = 2.

?- a(X,Y) = a(b(c,Y),Z).

X = b(c, Z),

Y = Z.

?- tree(left, root, Right) = tree(left, root, tree(a, b, tree(c, d, e))).

Right = tree(a, b, tree(c, d, e)).

QUIZ 2

1. What are the four items necessary for a PEAS description?

Performance, Environment, Actuators, Sensors

2. What is the evaluation of a search strategy while measuring performance of searching?

Completeness:

Can a solution eventually be found(Guaranteed)
Repeated states / loops / cycles can be avoided

Optimality:

Is the found solution the best among many solutions

Time complexity:

how long does it take to find a solution **Space complexity:**

how much memory is needed to perform the search

3. Please fill in the white squares (LXX) in the table below, and the words to be selected are

L1	L2	L3	L4	L5	
L6		L7		L8	
L9	L10	L11	L12	L13	L14
L15				L16	

word(d,o,g).
word(r,u,n).
word(t,o,p).
word(f,i,v,e).
word(l,o,s,t).
word(m,e,s,s).
word(u,n,i,t).
word(f,o,r,u,m).
word(g,r,e,e,n).
word(s,u,p,e,r).
word(v,a,n,i,s,h).
word(v,o,n,d,e,r).
word(y,e,l,l,o,w).

Try to write a rule solution.

solution(L1,L2,L3,L4,L5,L6,L7,L8,L9,L10,L11,L12,L13,L14,L15,L16):word(L1,L2,L3,L4,L5), word(L9,L10,L11,L12,L13,L14), word(L1,L6,L9,L15), word(L3,L7,L11), word(L5,L8,L13,L16).

1. Given a data set S, as it shows in the table below. Please calculate the *information gain* of 'outlook': Gain(S,Outlook)=?.

*Please give the necessary calculation steps.

outlook	temperature	humidity	wind	playTennis
sunny	hot	high	weak	no
sunny	hot	high	strong	no
overcast	hot	high	weak	yes
rain	mild	high	weak	yes
rain	cool	normal	weak	yes
rain	cool	normal	strong	no
overcast	cool	normal	strong	yes
sunny	mild	high	weak	no
sunny	cool	normal	weak	yes
rain	mild	normal	weak	yes
sunny	mild	normal	strong	yes
overcast	mild	high	strong	yes
overcast	hot	normal	weak	yes
rain	mild	high	strong	no

$$Entropy(S) = -p_{+} \log_{2} p_{+} - p_{-} \log_{2} p_{-}$$

$$Gain(S, A) = Entropy(S) - \sum_{v \in Values(A)} \frac{|S_v|}{|S|} Entropy(S_v)$$

1. Entropy(S) =
$$-(9/14) \log 2 (9/14)$$

- $(5/14) \log 2 (5/14)$
= 0.94

2. Let 'Outlook' as the root node, we have:

Entropy(Sunny)=
$$(-3/5)*log2(3/5)-(2/5)*log2(2/5)$$

=0.971;

3. Gain(Outlook)=0.940-(5/14)*Entropy(Sunny)

-(4/14)*Entropy(Overcast)

-(5/14)*Entropy(Rain)

=0.247

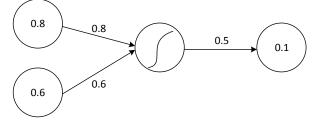
More details please find:

http://axon.cs.byu.edu/~martinez/classes/4 78/stuff/labhints/decisionTreeHelp.html 2. Given a table as below, please calculate the 1st and 2nd iteration for the weights using the data from the first row of the table according to the BP (Backpropagation) neural network.

Active function(sigmoid): $f(x) = \frac{1}{1 + e^{-x}}$

Learning rate: 0.3

X1	X2	Υ
0.8	0.6	0.1
0.5	-0.1	0.7
:	:	:



Input layer

Hidden layer

Output layer

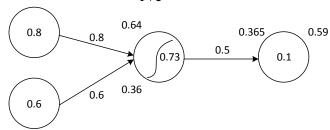
1. Calculate the value of hidden and output layer

O(hidden input)=0.8*0.8+0.6*0.6=1

O(hidden output)=
$$\frac{1}{1+e^{-1}}$$
 =0.73

O(output input)=0.73*0.5=0.365

O(output output)=
$$\frac{1}{1+e^{-0.365}}$$
 =0.59

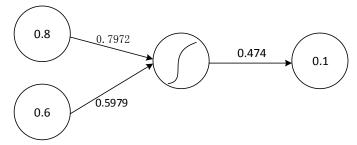


2. Error calculation for the 1st iteration

$$\begin{split} &\delta_{\mathbf{k}}\!=\!\mathbf{o}_{\mathbf{k}}(1\!-\!\mathbf{o}_{\mathbf{k}})(t_{k}-\mathbf{o}_{\mathbf{k}})\\ &=\!0.59^{*}(1\!-\!0.59)^{*}(0.1\!-\!0.59)\!=\!\!-0.1185\\ &\delta_{\mathbf{h}}\!=\!\mathbf{o}_{h}(1\!-\!\mathbf{o}_{h})\sum_{k\in outputs}w_{kh}\delta_{k}\\ &=\!0.73^{*}(1\!-\!0.73)^{*}0.5^{*}(-0.12)\!=\!-0.01168\\ &\Delta\mathbf{w}_{\mathbf{o}}\!=\!\eta\delta_{\mathbf{k}}\mathbf{x}_{\mathbf{h}}=\!0.3^{*}(-0.1185)^{*}0.73\!=\!\!-0.026\\ &\Delta\mathbf{w}_{11}\!=\!\eta\delta_{\mathbf{h}}\mathbf{x}_{11}\!=\!0.3^{*}(-0.01168)^{*}0.8\!=\!\!-0.0028\\ &\Delta\mathbf{w}_{21}\!=\!\eta\delta_{\mathbf{h}}\mathbf{x}_{21}\!=\!0.3^{*}(-0.01168)^{*}0.6\!=\!\!-0.0021 \end{split}$$

$$w_o = w_o + \Delta w_o = 0.5 + (-0.026) = 0.474$$

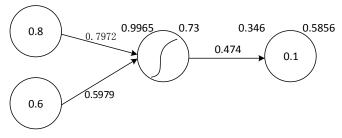
 $w_{11} = w_{11} + \Delta w_{11} = 0.8 + (-0.0028) = 0.7972$
 $w_{21} = w_{21} + \Delta w_{21} = 0.6 + (-0.0021) = 0.5979$



 Calculate the value of hidden and output layer of 1st iteration

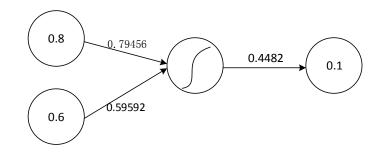
O(hidden output)=
$$\frac{1}{1+e^{-0.9965}}$$
 =0.73

O(output output)=
$$\frac{1}{1+e^{-0.346}}$$
 = 0.5856



4. Error calculation for the 2st iteration

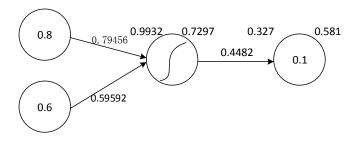
$$\begin{split} &\delta_{\mathbf{k}}\!=\!\mathbf{o}_{\mathbf{k}}(1\!-\!\mathbf{o}_{\mathbf{k}})(t_{k}-\mathbf{o}_{\mathbf{k}})\\ &=\!0.5856^{*}(1\!-\!0.5856)^{*}(0.1\!-\!0.5856)\!=\!-0.1178\\ &\delta_{\mathbf{h}}\!=\!\mathbf{o}_{h}(1\!-\!\mathbf{o}_{h})\sum_{k\in outputs}w_{kh}\delta_{k}\\ &=\!0.73^{*}(1\!-\!0.73)^{*}0.474^{*}(\!-\!0.1178)\!=\!-0.0110\\ &\Delta\mathbf{w}_{o}\!=\!\eta\delta_{\mathbf{k}}\mathbf{x}_{\mathbf{h}}=\!0.3^{*}(\!-\!0.1178)^{*}0.73\!=\!-0.0258\\ &\Delta\mathbf{w}_{11}\!=\!\eta\delta_{\mathbf{h}}\mathbf{x}_{11}\!=\!0.3^{*}(\!-\!0.0110)^{*}0.8\!=\!-0.00264\\ &\Delta\mathbf{w}_{21}\!=\!\eta\delta_{\mathbf{h}}\mathbf{x}_{21}\!=\!0.3^{*}(\!-\!0.0110)^{*}0.6\!=\!-0.00198\\ &w_{o}\!=\!\mathbf{w}_{o}\!+\!\Delta\mathbf{w}_{o}\!=\!0.474\!+\!(\!-\!0.0258)\!=\!0.4482\\ &w_{11}\!=\!\mathbf{w}_{11}\!+\!\Delta\mathbf{w}_{11}\!=\!0.7972\!+\!(\!-\!0.00264)\!=\!0.79456\\ &w_{21}\!=\!\mathbf{w}_{21}\!+\!\Delta\mathbf{w}_{21}\!=\!0.5979\!+\!(\!-\!0.00198)\!=\!0.59592 \end{split}$$



 Calculate the value of hidden and output layer of 2nd iteration

O(hidden output)=
$$\frac{1}{1+e^{-0.9932}}$$
 =0.7297

O(output output)=
$$\frac{1}{1+e^{-0.327}}$$
 = 0.581



This one hidden layer network is very simple just for easy calculation by hand, following shows another sample of a normal BP neural network to solve the XOR problem, which has 3 input layer, 4 hidden layer and one output layer coded with Python.

small change in any weight (or bias)

causes a small change in the output $\longrightarrow \text{output} + \Delta \text{output}$

BP_XOR.py:

```
# -*- codina: utf-8 -*-
                                                                      update()#updata weights
                                                                      if i%500==0:
                                                                        L1 = sigmoid(np.dot(X,V))#output of hidden
    import numpy as np
                                                               layer(4,4)
    #Input data
                                                                        L2 = sigmoid(np.dot(L1,W))#output of output
                                                               layer(4,1)
    X = np.array([[1,0,0],
                                                                        print('Current Error:',np.mean(np.abs(Y.T-
             [1,0,1],
                                                               L2)))
             [1,1,0],
                                                                   L1 = sigmoid(np.dot(X,V))#output of hidden
             [1,1,1]])
    #label
                                                               layer(4,4)
    Y = np.array([[0,1,1,0]])
                                                                   L2 = sigmoid(np.dot(L1,W))#output of output
    #Initial weights, [-1,1]
                                                               layer(4,1)
                                                                   print(L2)
    V = np.random.random((3,4))*2-1
    W = np.random.random((4,1))*2-1
                                                                   def judge(x):
    print(V)
                                                                      if x > = 0.5:
    print(W)
                                                                        return 1
    #learning rate
    Ir = 0.11
                                                                      else:
                                                                         return 0
    def sigmoid(x):
                                                                   for i in map(judge,L2):
       return 1/(1+np.exp(-x))
                                                                      print(i)
    #derivation
    def dsigmoid(x):
                                                                      #test a new data
       return x*(1-x)
                                                                   X = np.array([[1,0,1],
                                                                            [1,1,1],
    def update():
       global X,Y,W,V,Ir
                                                                            [1,1,0],
                                                                            [1,0,0]])
    #Input,output,weights,weights,learn rate
       #L1: Input -->hidden; 3X4
                                                                   L1 = sigmoid(np.dot(X,V))#output of hidden
       # L2: hidden--> output; 1
                                                               layer(4,4)
       L1 = sigmoid(np.dot(X,V))#output of hidden
                                                                   L2 = sigmoid(np.dot(L1,W))#output of output
layer(4,4)
                                                               layer(4,1)
       L2 = sigmoid(np.dot(L1,W))#output of output
layer(4,1)
                                                                   def judge(x):
                                                                      if x > = 0.5:
       # L2_delta: error of output
                                                                        return 1
       #L1 delta: error of hidden
                                                                      else:
       L2 delta = (Y.T - L2)*dsigmoid(L2)
                                                                        return 0
       L1_delta = L2_delta.dot(W.T)*dsigmoid(L1)
                                                                   for i in map(judge,L2):
       #W C: weights changing from input to
                                                                 print(i)
hidden layer
       # V_C: weights changing from hidden to
output layer
       W C = Ir^*L1.T.dot(L2 delta)
       V_C = Ir^*X.T.dot(L1_delta)
       W = W + W C
       V = V + V C
    for i in range(20000):
```