

## Exercise 6 - Self Organising Map (in R)

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First model map output:

x	x1	x2	x3	x4	x5	x6	x7	x8	x9	x10
1	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
2	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
3	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
4	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
5	XXXXXXXXXX	Cow	Horse+Zebra	XXXXXXXXXX	XXXXXXXXXX	Hawk+Owl	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
6	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Eagle	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Goose+Duck
7	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Dove
8	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Hen	XXXXXXXXXX	XXXXXXXXXX
9	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Dog+Wolf	XXXXXXXXXX	XXXXXXXXXX
10	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Lion+Tiger	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Cat+Fox

Figure 1 - Output map from SOM algorithm

First model human-clustered map:

Legend: O - large mammals, O - prey birds, O - small birds, O - prey mammals

x	x1	x2	x3	x4	x5	x6	x7	x8	x9	x10
1	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
2	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
3	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
4	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
5	XXXXXXXXXX	Cow	Horse+Zebra	XXXXXXXXXX	XXXXXXXXXX	Hawk+Owl	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
6	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Eagle	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Goose+Duck
7	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Dove
8	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Hen	XXXXXXXXXX	XXXXXXXXXX
9	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Dog+Wolf	XXXXXXXXXX	XXXXXXXXXX
10	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Lion+Tiger	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Cat+Fox

Figure 2 - Manually annotated map from SOM algorithm

Second model map output:

x	x1	x2	x3	x4	x5	x6	x7	x8	x9	x10
1	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Tiger	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Cat
2	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
3	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
4	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Fox	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
5	Eagle	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
6	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Cow
7	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Owl+Hawk	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Lion	XXXXXXXXXX	XXXXXXXXXX
8	XXXXXXXXXX	Dove	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	Wolf	XXXXXXXXXX	XXXXXXXXXX
9	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX	XXXXXXXXXX
10	Hen	XXXXXXXXXX	XXXXXXXXXX	Duck+Goose	XXXXXXXXXX	XXXXXXXXXX	Dog	XXXXXXXXXX	XXXXXXXXXX	Horse+Zebra

Figure 3 - Output map from refined SOM algorithm

Final model map output:

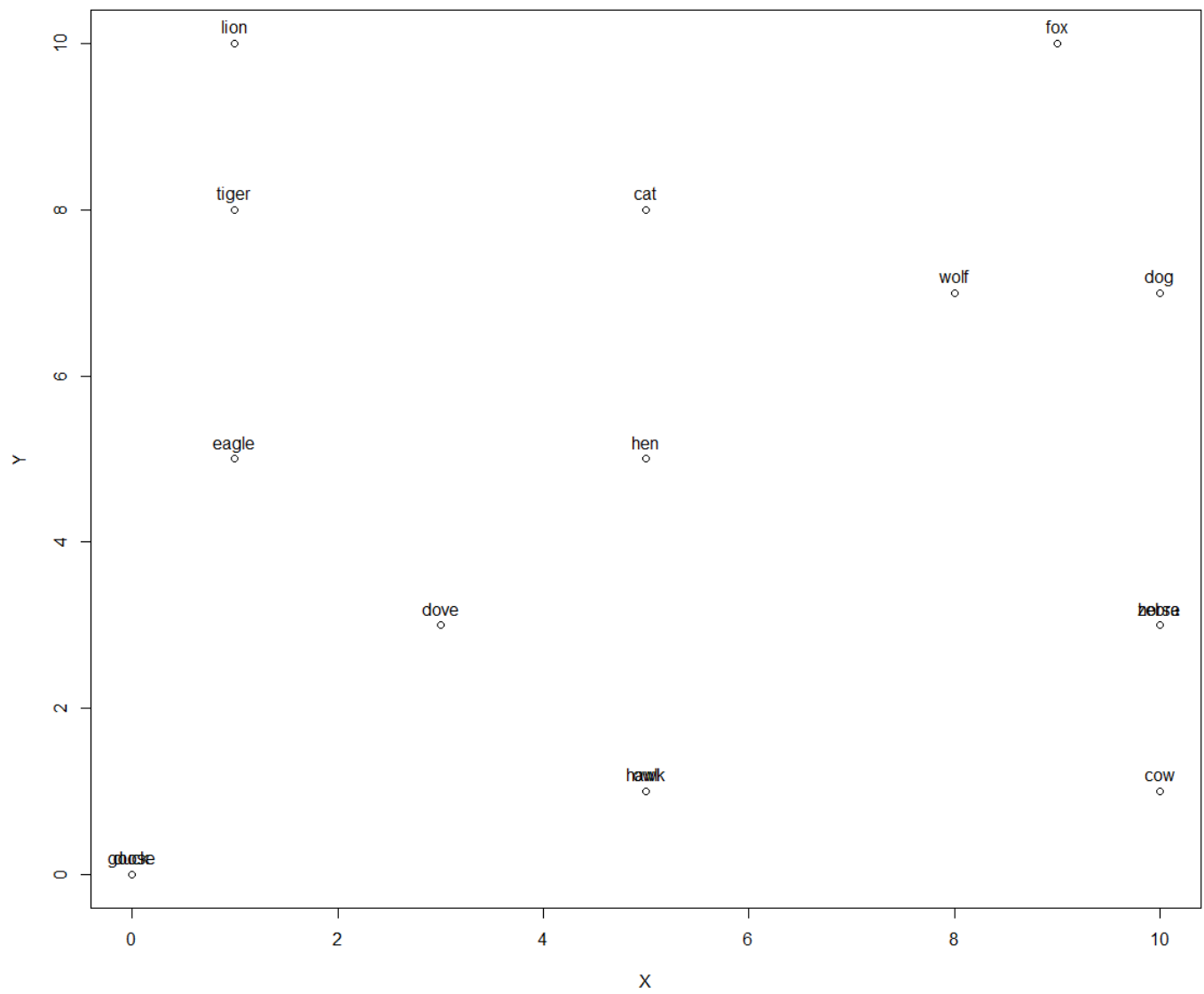


Figure 4 - Output map from final SOM algorithm

Script:

#Author: Thomas Hollis  
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#1. Data Import

```
data <- read.csv("R/AnimalsSOMnolab.csv")
```

#2. Global variables

```
t <- 1  
d <- matrix(0, 10, 10)  
pattern_size <- 13  
map <- matrix(list(), 10, 10)  
positionx <- integer(16)  
positiony <- integer(16)  
SOM <- matrix(0,2,16)  
colnames(SOM) <- c("dove", "hen", "duck", "goose", "owl", "hawk", "eagle", "fox",  
"dog", "wolf", "cat", "tiger", "lion", "horse", "zebra", "cow")  
rownames(SOM) <- c("X", "Y")
```

#3. Set random map weights

```
for (i in 1:10)  
{  
  for (j in 1:10)  
  {  
    map[[i,j]] <- runif(13, 0, 1)  
  }  
}
```

#4. Pick animal at random

```
for (t in 1:10000)  
{  
  r <- round(runif(1, 1, 16))  
  x <- data[r,]  
  X <- t(x)  
  for (i in 1:10)  
  {  
    for (j in 1:10)  
    {  
      w <- map[[i,j]]  
      W <- t(w)  
      d[i,j] <- sqrt((as.numeric(x)-as.numeric(W))%*%(as.numeric(X)-as.numeric(w)));  
    }  
  }  
  distance <- d[1,1]  
  for (i in 1:10)  
  {  
    for (j in 1:10)  
    {  
      if (d[i,j] < distance)  
      {  
        distance <- d[i,j]  
        u <- i  
        v <- j  
      }  
    }  
  }  
}
```

```

}
#setting learning parameters
theta <- 1;
alpha <- 100/(200+t);
for (i in 1:10)
{
  for (j in 1:10)
  {
    eta <- exp(-(((i-u)^2)+(j-v)^2)/2*theta^2)
    map[[i,j]] <- map[[i,j]] + alpha*eta*(X-map[[i,j]])
  }
}
}

#TESTING PHASE
for (test in 1:16)
{
  y <- data[test,]
  Y <- t(y)
  for (i in 1:10)
  {
    for (j in 1:10)
    {
      w <- map[[i,j]]
      W <- t(w)
      d[i,j] <- sqrt((as.numeric(y)-as.numeric(W))%*%(as.numeric(Y)-as.numeric(w)))
    }
  }
  distance <- d[1,1]
  for (i in 1:10)
  {
    for (j in 1:10)
    {
      if (d[i,j] < distance)
      {
        distance <- d[i,j]
        positionx[test] <- i
        positiony[test] <- j
      }
    }
  }
}

#Mapping each animal
for(i in 1:16)
{
  SOM[1,i] <- positionx[i]
  SOM[2,i] <- positiony[i]
}

plot(t(SOM))
text(t(SOM), labels = colnames(SOM), pos = 3)

```