

Entropy Based Reduct (EBR) Computation Using Sequential Backward Elimination(SBE) Technique

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Introduction:-

In this reduct computation technique we compute the Reduct of the given Decision System based on the entropy calculation, developed from work carried out in Jensen and Shen (2004a). EBR is concerned with examining a dataset and determining those attributes that provide the most gain in information. The entropy of attribute A (which can take values a_1, \dots, a_m) with respect to the conclusion C (of possible values c_1, \dots, c_n) is defined as: follows :-

$$H(C | A) = - \sum_{j=1}^m p(a_j) \sum_{i=1}^n p(c_i | a_j) \log_2 p(c_i | a_j)$$

Where:-

m= is number of unique objects with respect to selected set of attribute, and
n=number of unique decision with respect to decision attribute

Assumptions:-

In our implementation we assume that

1. Last Attribute of given D.S. is decision attribute
2. System is consistent

Language Used:-

R Language

Function Implemented:-

Two functions are implemented named as

- 1.reduct()
- 2.entropy()

1.reduct():-

This function compute the reduct set based on the Sequential Backward Elimination.

Initially we consider **entire set of conditional attribute as reduct set** and compute the **entropy** for that set with respect to decision attribute by calling **entropy()** function. We called this entropy value as “**oldentrp**” in our reduct function, and if “**oldentrp>0**” we simply print that “unable to find out reduct because initial

entropy is greater than 0".

when "**oldentrp==0**" then we enter into the loop to perform actual reduct computation where we first compute the **entropy for each individual attribute**(to decide which attribute to be selected first to remove from initial reduct set i.e from all conditional set). and store it into array of attribute where index represent attribute **ent_redAtr[i]**.

We remove the attribute having highest entropy value among all conditional attribute and then we again find out the entropy of **new reduct set(T1)**.

if "**newEntropy==0**" then we continue this process (repeat) until we have number of **conditional attribute greater than or equal to 1**(total attribute 2 ,one conditional and one decision) or "**newEntropy==0**".

in each iteration at the end if that attribute is successful removed from reduct set we make that attribute entropy equal to 0 in **ent_redAtr[i]** so that we can select next attribute having highest attribute from remaining.

when "**total attribute(tA)<2**" or "**newEntropy>0**" we break the loop and come out side the loop.

```
entropy.R x  reductComp.R x
Source on Save
1 reduct=function()
2 {
3   dectab=read.csv("input.csv") #file name to read data
4   print("Decision System :")
5   print(dectab)
6   tA=ncol(dectab) #total number of attributes
7   cA=tA-1 #conditionl attribute
8   oldentrp=entropy(tA,dectab) #calculate for entire conditional Atri Set
9   print("Entropy Of All Cond Attribute Set:")
10  print(oldentrp)
11  reduct_set=c(1:cA) #initial reduct set
12  ent_redAtr=c(1:cA) #to hold value of individual Cond attribute
13  if(oldentrp==0) #DO tis only if initial entropy of all attribute is equal to 0
14  {
15    for(i in 1:cA)
16    {
17      reduct_set[i]=i #INITIALIZATION OF INITIAL REDUCT SET
18      ent_redAtr[i]=entropy(2,dectab[,c(i,tA)]) #Compute Entropy Of Individual Cond. Attribute
19    }
20    print("Entropy Of Individual Attribute::")
21    print(ent_redAtr)
22    print("Initial Reduct Set::")
23    print(reduct_set)
24    repeat #REDUCT COMPUTAION ALGORITHEM IMPLEMENTATION
25    {
26      val=which.max(ent_redAtr)
27      print("Attribute Selected to remove::")
28      print(val)
29      T1=setdiff(reduct_set, val) #set excluding selected attribute
30      tA1=length(T1)+1
31      T2=union(T1,tA)
32      dectab1=dectab[,T2]
33      if(tA1>=2)
34      {
35        print("New Decision System::")
36        print(cbind(dectab1))
37        newEntropy=entropy(tA1,dectab1) #COMPUT ENTROPY OF NEW REDUCT SET
```

```

37     newEntropy=entropy(tA1,dectab1) #COMPUT ENTROPY OF NEW REDUCT SET
38     print("Entropy::")
39     print(newEntropy)
40     if(newEntropy==0)
41     {
42         reduct_set=T1
43         print("New Reduct Is:")
44         print(reduct_set)
45         ent_redAtr[val]=0 #change the entropy of selected attribute to 0
46     }
47     else
48     {
49         print("Imp Attribute for Reduct Can't Remove::")
50         print(val)
51         T1=union(reduct_set, val) #add that attribute back to reduct Set
52         break
53     }
54 }
55 else
56 {
57     print("Can't Remove This Attribute,Only 1 Cond Attribute left::")
58     T1=union(reduct_set, val)
59     break #when only one conditional attribute left
60 }
61 reduct_set=T1
62 }
63 reduct_set=T1
64 print("Final Reduct Is:")
65 print(reduct_set)
66 }
67 else # when all cond attribute set entropy >0
68 {
69     print("No reduct for This Dec System,Initial Entropy of All Cond.Attribute >0::")
70     print(oldentrp)
71 }
72 }

```

Note:-

finding out the entropy of individual attribute and removing highest entropy attribute from reduct set saves lot of computation time.

2.entropy() :-

Take two arguments **1st total number of attributes(tA1)** and **2nd dec system(dectab1)** for which we want to compute the entropy.

In this function **table1= cbind(dsctab[, -dA])** is use to convert the vacter into row wise when onle single attribute left in conditional attribute set and -dA is used to exclude the decsion attribute from D.S to create onlt conditioanl attribute sysetm.

“**uniquetable**” is conditional attribute table with unique object entry for this **unique(table1)** function is used.

“**val1**” or (**aj**) one object of unique cond attribute table i.e object of “**uniquetable**”. “**val2**” object of normal conditioanl attribute table. Then compare val1 and val2 to find out how many time val1 repeated in cond attribute table to find out probability of (**aj**). And increase the counter (**count**) if both are same.

```

1 entropy=function(dA,dsctab)
2 {
3   cA=dA-1 #number of conditional attribute
4   table1=cbind(dsctab[, -dA]) #to convert vector into row
5   uniqtable=unique(table1) #Unique Object table of cond System
6   m=0 # counter
7   count2=0 #to find out probability of each element in 2nd fun
8   manprob=0
9   entp1=0
10  for(n in 1:nrow(uniqtable))
11  {
12    count=0
13    val1=uniqtable[n,] #selcet each unique object
14    for(y in 1:nrow(table1))#loop to find out probability of individual unique object(p(aj))
15    {
16      val2=table1[y,]
17      if(all(val1==table1[y,]))# compare val1 with each object of cond table
18      {
19        count=count+1 #number of time val1 object repeate
20      }
21    }
22    unique_ds=unique(dsctab[,dA]) #find out total number of decision
23    sum1=0
24    for(y in 1:length(unique_ds)) #to find out all values p(Ci/aj)logp(Ci/aj) for each unique aj
25    {
26      count2=0
27      val3=unique_ds[y] # select one decision
28      unio=merge(val1,val3) # combine dec value with selected unique object
29      for(m in 1:nrow(dsctab))
30      {
31        if(all(dsctab[m,]==unio))
32        {
33          count2=count2+1 #number of time that unique object with decision repeate in dec table
34        }
35      }
36      if(count2!=0)
37      {
38        p2=(count2/count)*(log(count2/count)/log(2))#find prob value for that object
39        sum1=sum1+p2 # add to privious dec attribute value
40      }
41    }
42    entp1=entp1+(-(count/nrow(table1))*sum1) #final entropy
43  }
44  return(entp1)
45 }

```

And then we find out **Sumation of $P(c_i/a_j)\log(p(c_i/a_j))$** .

Where “**unique_ds**” is vector of unique decision attribute.

“**sum1**” is used to hold the sum of all values from (1 to m) for each decision w.r.t. **aj**

“**val3**” is selected **unique object(aj)**+ selected **Decision**

and for each val3 we find out number of time that will occure in decision system and increas the counter if match is found.

then we compute $p2\{\text{Sumation of } P(c_i/a_j)\log(p(c_i/a_j))\}.$

and add it to sum1.

Then add this sum1 value to the variable **entp1** which is used to store the **entropy value**

repeate this for all unique object of cond attribute table and finaly return the entropy value(entp1).

Results:-

Following is the results of the given decision system in the paper (**Chapter 3 Rough Set -Based Feature Selection:A review by Richard Jensen**)

In our example dataset, EBR first evaluates the entropy of initial reduct set {A1,A2,A3,A4},

Entropy

{A1,A2,A3,A4}= 0

if initial reduct set entropy==0, Then calculate entropy for each individual attribute:

Entropy

{A1} 1.1887219

{A2} 0.75

{A3} 0.9387219

{A4} 0.75

The Attribute with highest entropy here is {A1}. The algorithm selects attribute {A1} to remove from Initial reduct set {A1,A2,A3,A4}, So set become{A2,A3,A4} then calculate entropy for this subset:

Entropy

{A2,A3,A4}=0

Hence new reduct become {A2,A3,A4}.

Again select the attribute (from {A2,A3,A4}) having highest entropy and that is {A3}, The algorithm selects attribute {A3} to remove from current reduct set{A2,A3,A4}, So subset become{A2,A4} then calculate entropy for this subset:

Entropy

{A2,A4}=0

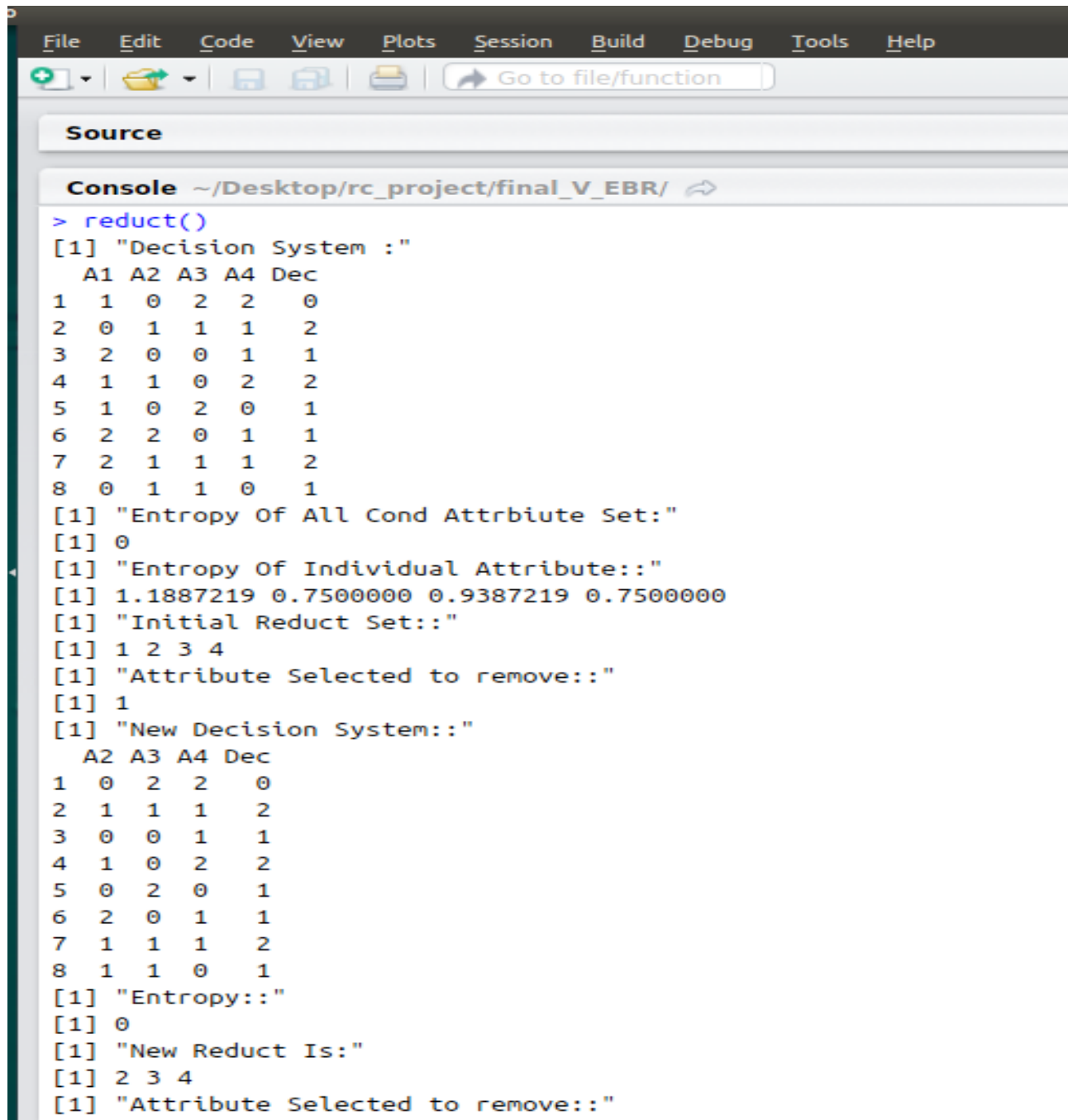
Hence new reduct become {A2,A4}.

Again select the attribute (from {A2,A4}) having highest entropy Here both A2 and A4 have same entropy 0.75, So algorithm selects attribute {A2} to remove from current reduct set{A2,A4}, So subset become{A4} then calculate entropy for this subset:

Entropy

{A4}>0

Stopping criterion has been met, add A2 back to the reduct set and return final Reduct as {A2,A4} as this value equals the entropy for the entire feature set ($H(D|\{A2,A4\}) = 0 = H(D|C)$). The algorithm terminates and returns this feature subset; the dataset can now be reduced to these features only. As the resulting entropy is zero, the returned subset is a rough set reduct.



```
> reduct()
[1] "Decision System :"
```

	A1	A2	A3	A4	Dec
1	1	0	2	2	0
2	0	1	1	1	2
3	2	0	0	1	1
4	1	1	0	2	2
5	1	0	2	0	1
6	2	2	0	1	1
7	2	1	1	1	2
8	0	1	1	0	1

```
[1] "Entropy Of All Cond Attrbiute Set:"
[1] 0
[1] "Entropy Of Individual Attribute::"
[1] 1.1887219 0.7500000 0.9387219 0.7500000
[1] "Initial Reduct Set::"
[1] 1 2 3 4
[1] "Attribute Selected to remove::"
[1] 1
[1] "New Decision System::"
```


	A2	A3	A4	Dec
1	0	2	2	0
2	1	1	1	2
3	0	0	1	1
4	1	0	2	2
5	0	2	0	1
6	2	0	1	1
7	1	1	1	2
8	1	1	0	1

```
[1] "Entropy::"
[1] 0
[1] "New Reduct Is:"
[1] 2 3 4
[1] "Attribute Selected to remove::"
```

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Source

Console ~/Desktop/rc_project/final_V_EBR/ 

```
[1] "New Reduct Is:"
[1] 2 3 4
[1] "Attribute Selected to remove::"
[1] 3
[1] "New Decision System::"
  A2 A4 Dec
1  0  2   0
2  1  1   2
3  0  1   1
4  1  2   2
5  0  0   1
6  2  1   1
7  1  1   2
8  1  0   1
[1] "Entropy::"
[1] 0
[1] "New Reduct Is:"
[1] 2 4
[1] "Attribute Selected to remove::"
[1] 2
[1] "New Decision System::"
  A4 Dec
1  2   0
2  1   2
3  1   1
4  2   2
5  0   1
6  1   1
7  1   2
8  0   1
[1] "Entropy::"
[1] 0.75
[1] "Imp Attribute for Reduct Can't Remove::"
[1] 2
[1] "Final Reduct Is:"
[1] 2 4
> |
```