

# Application of unsupervised clustering algorithm on a data set of Insulin like Growth Factor Receptor sequences

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**Abstract** - Clustering can be considered as the most unsupervised learning problem. A component being considered in a clustering algorithm is the distance measure between data points in the given set. In this work, we present a modified single and complete linkage clustering techniques were implemented in python language and applied on a dataset of IGF1R (Insulin like Growth Factor Receptor) protein sequences to generate orthologous clusters of sequences and the phylogenetic trees in each case were presented. From analysis, it was observed that IGF1R\_XENLA and PIG sequences formed one group in single linkage whereas IGF1R\_XENLA and RAT appeared under one clade in complete linkage algorithm.

**Index Terms:** Clustering, IGF1R, Insulin, Python, Receptors, Similarity, Sequencing.

## 1 INTRODUCTION

A cluster is a collection of objects which have similarities among other objects in the group and are probably dissimilar to the objects belonging to other clusters. Clustering can be considered the most important unsupervised learning problem. The goal of clustering is to determine the intrinsic grouping in a set of unlabeled data [1] [2]. Clustering algorithms may be classified as, Exclusive, overlapping, hierarchical and probabilistic clustering. In the first case data are grouped in an exclusive way, so that if a certain datum belongs to a definite cluster then it could not be included in another cluster [3]. The overlapping clustering uses fuzzy sets to cluster data, so that each point may belong to two or more clusters with different degrees of membership. A hierarchical clustering algorithm is based on the union between the two nearest clusters. Finally, the last kind of clustering uses a completely probabilistic approach [4] [5] [6]. An important component of a clustering algorithm is the distance measure between data points.

Human endothelial cells express IR, IGF-1 receptors (IGF-1R), and hybrid receptors (IR/IGF-1R) composed of heterodimers [7] [8]. IGF-1R are more abundant (10-fold higher) than IR [9]. In non endothelial cells, IGF-1R expression modulates insulin signalling by altering the levels of hybrid receptors [10]. It has been reported that type 2 diabetes and obesity are associated with increases in IR/IGF-1R expression in insulin-sensitive tissues [11]. In these pathological states, interventions aimed at down regulating IGF-1R expression may augment endothelial insulin sensitivity. Owing to this importance, in this paper, we report single and average linkage algorithms for a set of IGF1R sequences.

## 2 MATERIALS AND METHODS

### 2.1 Sequence Data

The IGF1R protein sequences were extracted from Swiss-Prot database [12] [13]. A data set of 6 IGF1R sequences is downloaded as fasta files. These are further subjected to multiple sequence analysis using clustalw [14] to generate similarity scores among all 6 sequences under study.

### 2.2 Hierarchical Clustering Algorithms

A similarity matrix is constructed for a given a set of  $N$  items:

1. Each item is added to the cluster there by creating  $N$  clusters, each containing just one item. Distances or similarities between the clusters are calculated.
2. The most similar clusters are paired and joined as a single cluster.
3. Now distances are computed between new cluster group and each of the old ones based on single and average linkage clustering. In single linkage, the distance between one cluster and another cluster is considered to be identical to the shortest distance from any object of one cluster to any object of the other cluster. In complete linkage, the distance between one cluster and another cluster is considered to be equal to the greatest distance from any object of one cluster to any object of the other cluster.
4. Repeat steps 2 and 3 until all items are clustered into a single cluster of size  $N$ .

## 3 RESULTS AND DISCUSSION

Insulin sequences are searched in Uniprot Knowledgebase. The database was accessed from Expert Protein Analysis System (ExPASy) server maintained by Swiss Institute of

Bioinformatics (SIB). It was reported in literature that *Insulin growth factor* and *Insulin* proteins compete to bind with Insulin receptor. Similarly *Insulin growth factor* protein competes to bind with *Insulin Growth Factor 1 Receptor*. An

I increased IGF-1R expression is associated with increased IR/IGF-1R and reduced numbers of IR holoreceptors [15].

The output of multiple alignments is given below.

```
sp|073798|IGF1R_XENLA      YRKVFENFLHNSIFVPRPNRRRDVLAVGNSTVTSYEKNSTTEDFSNFSDSERDDIEYPF 771
sp|P24062|IGF1R_RAT      YRKVFENFLHNSIFVPRPERRRRDVLQVANTTMSSRSRNTTVADTYNITDPEEFETEYPF 778
sp|Q60751|IGF1R_MOUSE    YRKVFENFLHNSIFVPRPERRRRDVMQVANTTMSSRSRNTTVADTYNITDPEEFETEYPF 778
sp|Q05688|IGF1R_BOVIN    -----NAIFVPRPERKRRREVMQIANTTMSSRSRNTTVLDTYNITDPEELETEYPF 50
sp|P08069|IGF1R_HUMAN    YRKVFENFLHNSIFVPRPERKRRRDVMQVANTTMSSRSRNTTAADTYNITDPEELETEYPF 777
sp|Q29000|IGF1R_PIG      -----0

sp|073798|IGF1R_XENLA    YETKVDYKWERTVISNLQPFTLYRIDIHSCNHEAEKLGCSASNFFVFARTMPAAGADDIPG 831
sp|P24062|IGF1R_RAT      FESRVDNK-ERTVISNLRPFTLYRIDIHSCNHEAEKLGCSASNFFVFARTMPAAGADDIPG 837
sp|Q60751|IGF1R_MOUSE    FESRVDNK-ERTVISNLRPFTLYRIDIHSCNHEAEKLGCSASNFFVFARTMPAAGADDIPG 837
sp|Q05688|IGF1R_BOVIN    FESRVDNK-ERTVISNLRPFTLYRIDIHSCNHEAEKLGCSASNFFVFARTMPAAGADDIPG 109
sp|P08069|IGF1R_HUMAN    FESRVDNK-ERTVISNLRPFTLYRIDIHSCNHEAEKLGCSASNFFVFARTMPAAGADDIPG 836
sp|Q29000|IGF1R_PIG      -----ERTVISNLRPFTLYRIDIHSCNHEAEKLGCSASNFFVFARTMPAAGADDIPG 51
                          *****:*****

sp|073798|IGF1R_XENLA    IVNTKEEDDGVIPLGWPPLRPNGLILMYEIEYKHQGEVHRECVSRQDYRKNGGIKLVRL 891
sp|P24062|IGF1R_RAT      PVTWEPRPENSIFLKWPEPENPNGLILMYEIKYGSQVEDQRECVSRQEYRKYGGAKLNRL 897
sp|Q60751|IGF1R_MOUSE    PVTWEPRPENSIFLKWPEPENPNGLILMYEIKYGSQVEDQRECVSRQEYRKYGGAKLNRL 897
sp|Q05688|IGF1R_BOVIN    PVTWEPRPENSIFLKWPEPENPNGLILMYEIKYGSQVEDQRECVSRQEYRKYGGAKLNRL 169
sp|P08069|IGF1R_HUMAN    PVTWEPRPENSIFLKWPEPENPNGLILMYEIKYGSQVEDQRECVSRQEYRKYGGAKLNRL 896
sp|Q29000|IGF1R_PIG      PVTWEPRPENSIFLKWPEPENPNGLILMYEIKYGSQVEDQRECVSRQEYRKYGGAKLNRL 111
                          *: : : ** * * * : * * : * * * * *

sp|073798|IGF1R_XENLA    PPGNYSAQVQAISLYNGSWTEPMSFCVKLKPQVRNINLQMVVAIPLALSFLLVGIISIV 951
sp|P24062|IGF1R_RAT      NPGNYTARIQATSLSGNGSWTDPVFFYVPAKTTY-ENFMHLIALPVAILLIVGGLVIML 956
sp|Q60751|IGF1R_MOUSE    NPGNYTARIQATSLSGNGSWTDPVFFYVPAKTTY-ENFMHLIALPVAILLIVGGLVIML 956
sp|Q05688|IGF1R_BOVIN    NPGNYTARIQATSLSGNGSWTDPVFFYVQAKTTY-ENFIHLMIALPIAVLLIVGGLVIML 228
sp|P08069|IGF1R_HUMAN    NPGNYTARIQATSLSGNGSWTDPVFFYVQAKTGY-ENFIHLIALPVAVLLIVGGLVIML 955
sp|Q29000|IGF1R_PIG      NPGNYTARIQATSLSGNGSWTEPFFYVQAKTTY-ENFIHLIALPVAVLLIVGGLVIML 170
                          ****:*:* ** * * * * * : * * * * * : * * * * *

sp|073798|IGF1R_XENLA    CFVFKKRNSNRLGNGVLYASVNPPEYFSAEMVVPDKWEVPREKITMNRLELGGQSFQGMVYE 1011
sp|P24062|IGF1R_RAT      YVFHRKRNSNRLGNGVLYASVNPPEYFSAADVVPDEWEVAREKITMNRLELGGQSFQGMVYE 1016
sp|Q60751|IGF1R_MOUSE    YVFHRKRNSNRLGNGVLYASVNPPEYFSAADVVPDEWEVAREKITMNRLELGGQSFQGMVYE 1016
sp|Q05688|IGF1R_BOVIN    YVFHRKRNSNRLGNGVLYASVNPPEYFSAADVVPDEWEVAREKITMSRELELGGQSFQGMVYE 288
sp|P08069|IGF1R_HUMAN    YVFHRKRNSNRLGNGVLYASVNPPEYFSAADVVPDEWEVAREKITMSRELELGGQSFQGMVYE 1015
sp|Q29000|IGF1R_PIG      YVFHRKRNSNRLGNGV-----186
                          ...:****.*****

sp|073798|IGF1R_XENLA    GIAKGVVKDEAETKVAIKTVNEAASMRERIEFLNEASVMKEFNCHHVRLLVVVSQGP 1071
sp|P24062|IGF1R_RAT      GVAKGVVKDEPETRVAIKTVNEAASMRERIEFLNEASVMKEFNCHHVRLLVVVSQGP 1076
sp|Q60751|IGF1R_MOUSE    GVAKGVVKDEPETRVAIKTVNEAASMRERIEFLNEASVMKEFNCHHVRLLVVVSQGP 1076
sp|Q05688|IGF1R_BOVIN    GVAKGVVKDEPETRVAIKTVNEAASMRERIEFLNEASVMKEFNCHHVRLLVVVSQGP 348
sp|P08069|IGF1R_HUMAN    GVAKGVVKDEPETRVAIKTVNEAASMRERIEFLNEASVMKEFNCHHVRLLVVVSQGP 1075
sp|Q29000|IGF1R_PIG      -----186
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sp|O73798|IGF1R_XENLA   LVIMELMTRGDLKSYLRSLRPDTEISNSG-QPTPSLKKMIQMAGEIADGMSYLNANKFVHR 1130
sp|P24062|IGF1R_RAT    LVIMELMTRGDLKSYLRSLRPEVE-NNLVLIPPSLSKMIQMAGEIADGMAYLNANKFVHR 1135
sp|Q60751|IGF1R_MOUSE  LVIMELMTRGDLKSYLRSLRPEVEQNNLVIPPSLSKMIQMAGEIADGMAYLNANKFVHR 1136
sp|Q05688|IGF1R_BOVIN  LVIMELMTRGDLKSYLRSLRPEMEN-NPVLAPPSLSKMIQMAGEIADGMAYLNANKFVHR 407
sp|P08069|IGF1R_HUMAN  LVIMELMTRGDLKSYLRSLRPEMEN-NPVLAPPSLSKMIQMAGEIADGMAYLNANKFVHR 1134
sp|Q29000|IGF1R_PIG    ----- 186

sp|O73798|IGF1R_XENLA   DLAARNCMVTEDFTVKIGDFGMDRDYIYTDYYRKGKGLLPVRWMSPELKDGVFTTNSD 1190
sp|P24062|IGF1R_RAT    DLAARNCMVAEDFTVKIGDFGMDRDYIYTDYYRKGKGLLPVRWMSPELKDGVFTTHSD 1195
sp|Q60751|IGF1R_MOUSE  DLAARNCMVAEDFTVKIGDFGMDRDYIYTDYYRKGKGLLPVRWMSPELKDGVFTTHSD 1196
sp|Q05688|IGF1R_BOVIN  DLAARNCMVAEDFTVKIGDFGMDRDYIYTDYYRKGKGLLPVRWMSPELKDGVFTTHSD 467
sp|P08069|IGF1R_HUMAN  DLAARNCMVAEDFTVKIGDFGMDRDYIYTDYYRKGKGLLPVRWMSPELKDGVFTTYS 1194
sp|Q29000|IGF1R_PIG    ----- 186

sp|O73798|IGF1R_XENLA   VWSFGVVLWEIATLAEQPYQGLSNEQVLRFMVEGGLLDKPDNCPDMLFELMRMCWQYNPK 1250
sp|P24062|IGF1R_RAT    VWSFGVVLWEIATLAEQPYQGLSNEQVLRFMVEGGLLDKPDNCPDMLFELMRMCWQYNPK 1255
sp|Q60751|IGF1R_MOUSE  VWSFGVVLWEIATLAEQPYQGLSNEQVLRFMVEGGLLDKPDNCPDMLFELMRMCWQYNPK 1256
sp|Q05688|IGF1R_BOVIN  VWSFGVVLWEIATLAEQPYQGLSNEQVLRFMVEGGLLDKPDNCPDMLFELMRMCWQYNPK 527
sp|P08069|IGF1R_HUMAN  VWSFGVVLWEIATLAEQPYQGLSNEQVLRFMVEGGLLDKPDNCPDMLFELMRMCWQYNPK 1254
sp|Q29000|IGF1R_PIG    -----MLFELMRMCWQYNPK 201
                                     *****:***

sp|O73798|IGF1R_XENLA   MRPSFLEIISIKDELDPGFKEVSFFYSEENKPPDTEELD----LEAENMESIPLDPSCA 1306
sp|P24062|IGF1R_RAT    MRPSFLEIIGSIKDEMEPSFQEVSFYSEENKPPPEELEMELELEPENMESVPLDPSAS 1315
sp|Q60751|IGF1R_MOUSE  MRPSFLEIIGSIKDEMEPSFQEVSFYSEENKPPPEELEMELELEPENMESVPLDPSAS 1316
sp|Q05688|IGF1R_BOVIN  MRPSFLEIISSVKDEMEAGFREVSFFYSEENKPPPEEL----DLEPENMESVPLDPSAS 583
sp|P08069|IGF1R_HUMAN  MRPSFLEIISIKKEEENPGFREVSFFYSEENKLPPEEL----DLEPENMESVPLDPSAS 1310
sp|Q29000|IGF1R_PIG    MRPSFLEIISIKDEMEPGFREVSFFYSEENKPPPEEL----DLEPENMESVPLDPSAS 257
*****.:*:*:*.:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:

sp|O73798|IGF1R_XENLA   ---LQNSEHHAGHKSENG--PGVVLRASFDERQPYAHMNGGRKNERALPLPQSAC 1358
sp|P24062|IGF1R_RAT    SASLPLPERHSGHKAENG--PGVLVLRASFDERQPYAHMNGGRANERALPLPQSSTC 1370
sp|Q60751|IGF1R_MOUSE  SASLPLPERHSGHKAENGGPGVVLVLRASFDERQPYAHMNGGRANERALPLPQSSTC 1373
sp|Q05688|IGF1R_BOVIN  SASLPLPDRHSGHKAENGGPGVVLVLRASFDERQPYAHMNGGRKNERALPLPQSSTC 640
sp|P08069|IGF1R_HUMAN  SSSLPLPDRHSGHKAENGGPGVVLVLRASFDERQPYAHMNGGRKNERALPLPQSSTC 1367
sp|Q29000|IGF1R_PIG    SSSLPLPDRHSGHKAENGGPGVVLVLRASFDERQPYAHMNGGRKNER----- 304
* :*:***:** *:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:*:

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The summary file output of clustalw contains all the necessary that is required to perform clustering. The pair wise scores generated by the program are given as input to

the python based single linkage and average linkage clustering algorithm.

### Percent Identity Matrix - created by Clustal2.1

1:	sp O73798 IGF1R_XENLA	100.00	75.02	74.82	77.88	75.67	71.48
2:	sp P24062 IGF1R_RAT	75.02	100.00	99.71	96.08	96.04	96.36
3:	sp Q60751 IGF1R_MOUSE	74.82	99.71	100.00	95.94	95.98	96.05
4:	sp Q05688 IGF1R_BOVIN	77.88	96.08	95.94	100.00	97.66	97.70
5:	sp P08069 IGF1R_HUMAN	75.67	96.04	95.98	97.66	100.00	98.68
6:	sp Q29000 IGF1R_PIG	71.48	96.36	96.05	97.70	98.68	100.00

### 3.1 Single linkage clustering

The scores output from clustalw analysis were given as input to python program to cluster sequences based on single linkage algorithm. The output is given below.

Distance Calculation Method: \*\*\*\*\* Single Linkage Clustering \*\*\*\*\*

Objects:  
IGF1R\_XENLA  
IGF1R\_RAT  
IGF1R\_MOUSE  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
IGF1R\_PIG  
0 75.02 74.82 77.88 75.67 71.48  
0 99.71 96.08 96.04 96.36  
0 95.94 95.98 96.05  
0 97.66 97.70  
0 98.68  
0

Objects:  
( 'IGF1R\_XENLA', 'IGF1R\_PIG', 71.48 )  
IGF1R\_RAT  
IGF1R\_MOUSE  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
0 75.02 74.82 77.88 75.67  
0 99.71 96.08 96.04  
0 95.94 95.98  
0 97.66  
0

Objects:  
( ( 'IGF1R\_XENLA', 'IGF1R\_PIG', 71.48 ),  
'IGF1R\_MOUSE', 74.82 )  
IGF1R\_RAT  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
0 75.02 77.88 75.67  
0 96.08 96.04  
0 97.66  
0

Objects:  
( ( ( 'IGF1R\_XENLA', 'IGF1R\_PIG', 71.48 ),  
'IGF1R\_MOUSE', 74.82 ), 'IGF1R\_RAT', 75.02 )  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
0 77.88 75.67  
0 97.66  
0

Objects:  
( ( ( ( 'IGF1R\_XENLA', 'IGF1R\_PIG', 71.48 ),  
'IGF1R\_MOUSE', 74.82 ), 'IGF1R\_RAT', 75.02 ),  
'IGF1R\_HUMAN', 75.67 )  
IGF1R\_BOVIN  
0 77.88  
0

Objects:

( ( ( ( 'IGF1R\_XENLA', 'IGF1R\_PIG', 71.48 ),  
'IGF1R\_MOUSE', 74.82 ), 'IGF1R\_RAT', 75.02 ),  
'IGF1R\_HUMAN', 75.67 ), 'IGF1R\_BOVIN', 77.88 )

### 3.2 Complete linkage clustering

The outcome of clusters generated by the average linkage clustering method resulted in the following.

Distance Calculation Method: \*\*\*\*\* Complete Linkage Clustering \*\*\*\*\*

MULTIPLE SEQUENCES: Insulin like Growth Factor-1 Receptor  
IGF1R\_XENLA  
IGF1R\_RAT  
IGF1R\_MOUSE  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
IGF1R\_PIG  
0 75.02 74.82 77.88 75.67 71.48  
0 99.71 96.08 96.04 96.36  
0 95.94 95.98 96.05  
0 97.66 97.70  
0 98.68  
0

MULTIPLE SEQUENCES: Insulin like Growth Factor-1 Receptor

( 'IGF1R\_XENLA', 'IGF1R\_RAT', 75.02 )  
IGF1R\_MOUSE  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
IGF1R\_PIG  
0 87.26 86.98 85.86 83.92  
0 95.94 95.98 96.05  
0 97.66 97.70  
0 98.68  
0

MULTIPLE SEQUENCES: Insulin like Growth Factor-1 Receptor

( ( 'IGF1R\_XENLA', 'IGF1R\_RAT', 75.02 ),  
'IGF1R\_MOUSE', 87.26499999999999 )  
IGF1R\_BOVIN  
IGF1R\_HUMAN  
IGF1R\_PIG  
0 91.46 90.92 89.98  
0 97.66 97.70  
0 98.68  
0

MULTIPLE SEQUENCES: Insulin like Growth Factor-1 Receptor

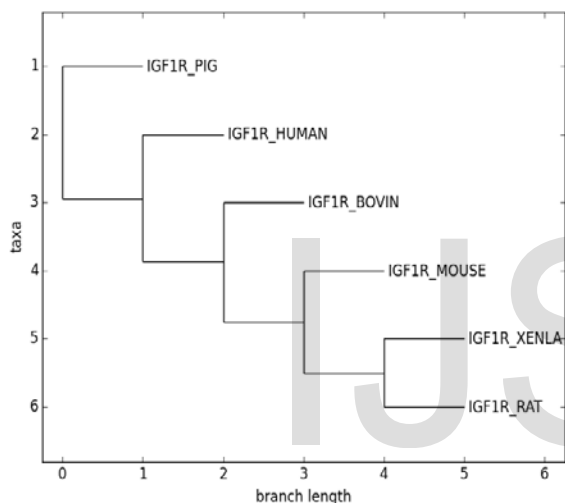
( ( ( 'IGF1R\_XENLA', 'IGF1R\_RAT', 75.02 ),  
'IGF1R\_MOUSE', 87.26499999999999 ),  
'IGF1R\_BOVIN', 91.46 )  
IGF1R\_HUMAN  
IGF1R\_PIG  
0 94.29 93.84  
0 98.68  
0

MULTIPLE SEQUENCES: Insulin like Growth Factor-1 Receptor

```
((('IGF1R_XENLA', 'IGF1R_RAT', 75.02),
'IGF1R_MOUSE', 87.26499999999999),
'IGF1R_BOVIN', 91.46), 'IGF1R_HUMAN',
94.28875)
IGF1R_PIG
0 96.26
0
```

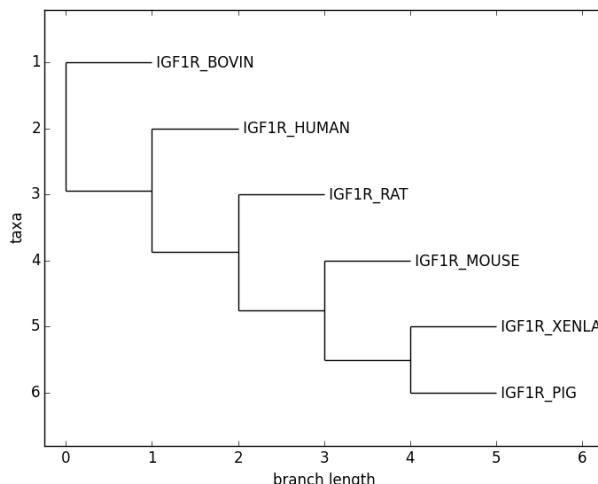
```
MULTIPLE SEQUENCES: Insulin like Growth
Factor-1 Receptor
((((('IGF1R_XENLA', 'IGF1R_RAT', 75.02),
'IGF1R_MOUSE', 87.26499999999999),
'IGF1R_BOVIN', 91.46), 'IGF1R_HUMAN',
94.28875), 'IGF1R_PIG', 96.26125)
```

Followed by the clusters generated by the single linkage and complete linkage analysis methods, phylogenetic trees were constructed using python code.



**Figure 1:** Phylogenetic tree obtained by single linkage method.

From Figure 1, it is evidenced that the single linkage clustering method clusters groups of similar sequences one after the other and the resultant phylogenetic tree can be viewed to understand the concept.



**Figure 2:** Phylogenetic tree obtained by complete linkage method.

From figures 1 and 2, it was observed that IGF1R\_XENLA and PIG sequences formed one group in single linkage whereas IGF1R\_XENLA and RAT appeared under one clade in complete linkage algorithm. Remaining sequences appeared as separate clades. Hence, in order to effectively cluster a set of IGF1R sequences, clustering by average linking technique was employed.

In our study, we employed Euclidean distance as distance measure. In this method, we considered the distance between one cluster and another cluster to be equal to the shortest distance from any member of one cluster to any member of the other cluster.

#### 4 CONCLUSION

Clustering techniques were implemented to study the relationship among IGF1R sequences based on the fact that analysis of gene data represents the important step towards identification of groups of genes that manifest similar expression pattern. Single linkage and average linkage clustering algorithms written in python language are used to generate clusters of orthologous insulin sequences, which resulted in a dendrogram. The pair wise scores generated from clustalw are given as input to the python based single linkage and average linkage clustering algorithm and the phylogenetic trees are reported.

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