

# Nature Inspired Visualization of Unstructured Big Data

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# Motivation

- Unstructured data is ubiquitous and is assumed to be around 80% of all data generated<sup>1</sup>
- Lack of recognizable structure and huge size makes it very challenging to work with Unstructured Large Datasets
- Classical Visualization – not suited for BigData; slow, memory hogging, limited dimensions

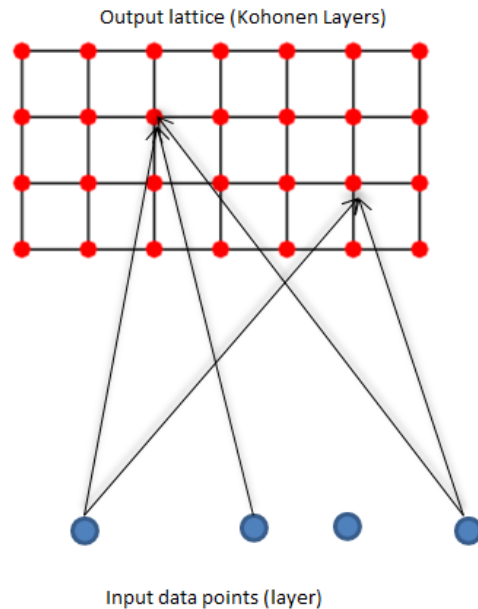
# Self Organizing Maps

- Unsupervised Machine Learning Technique
- Provides dimension reduction
- Multivariate Analysis
- Fast and low on memory (2D planar images)
- Reconstructing Self Organizing Maps as Spider Graphs for better visual interpretation

<sup>1</sup>Unstructured Data and the 80 Percent Rule, Clarabridge Bridgepoints, 2008 Q3. <http://clarabridge.com/default.aspx?tabid=137&ModuleID=635&ArticleID=551>

# Self Organizing Maps

- Artificial Neural Networks proposed by Teuvo Kohonen<sup>1</sup> which transforms the input dataset into two dimensional lattice
- Points in input layer are mapped onto 2D lattice, making each point potentially a Neuron



$$d_j(\mathbf{x}) = \sum_{i=1}^D (x_i - w_{ji})^2$$

Figure: Discriminant Function

where,

$x$  = point on Input Layer

$w$  = weight of the input point ( $x$ )

$i$  = all the input points

$j$  = all the neurons on the lattice

$d$  = Euclidean distance

Figure: Kohonen Network

<sup>1</sup>Kohonen, T.; "The self-organizing map," Proceedings of the IEEE , vol.78, no.9 <http://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=58325&isnumber=21>

# Current Visualization of SOM

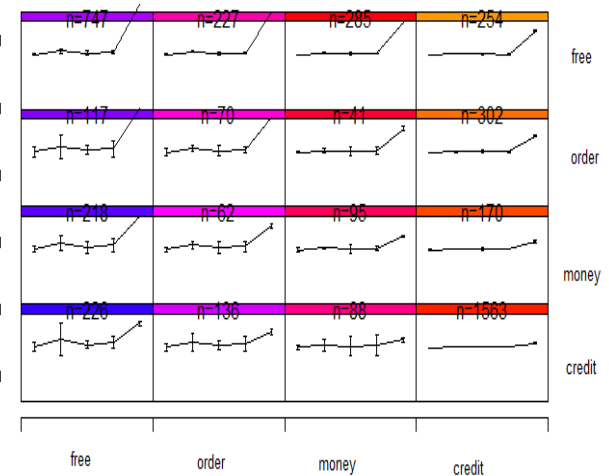
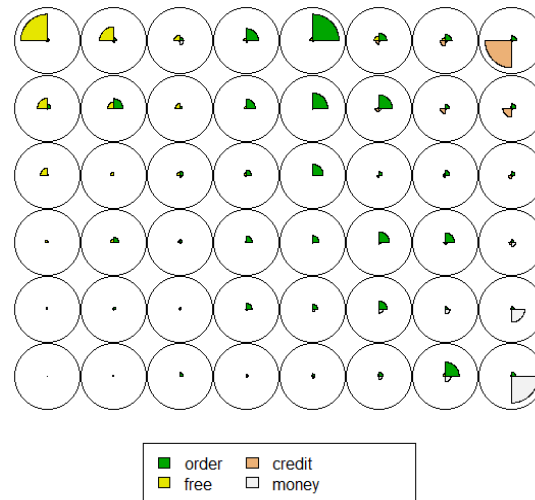
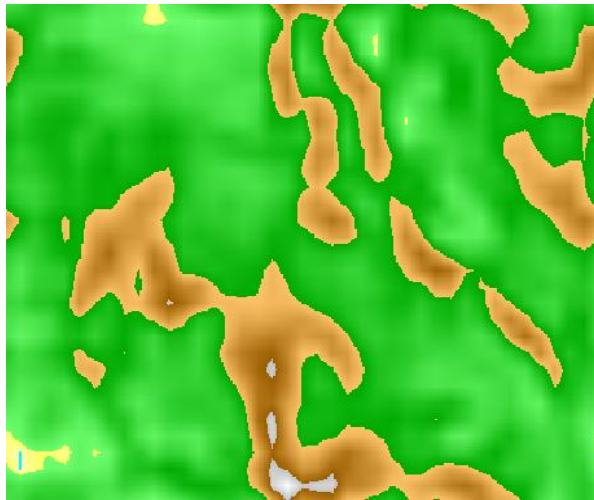


Fig: RapidMiner Tool (AGPL)  
Shows the Kohonen Map obtained after training the Neurons

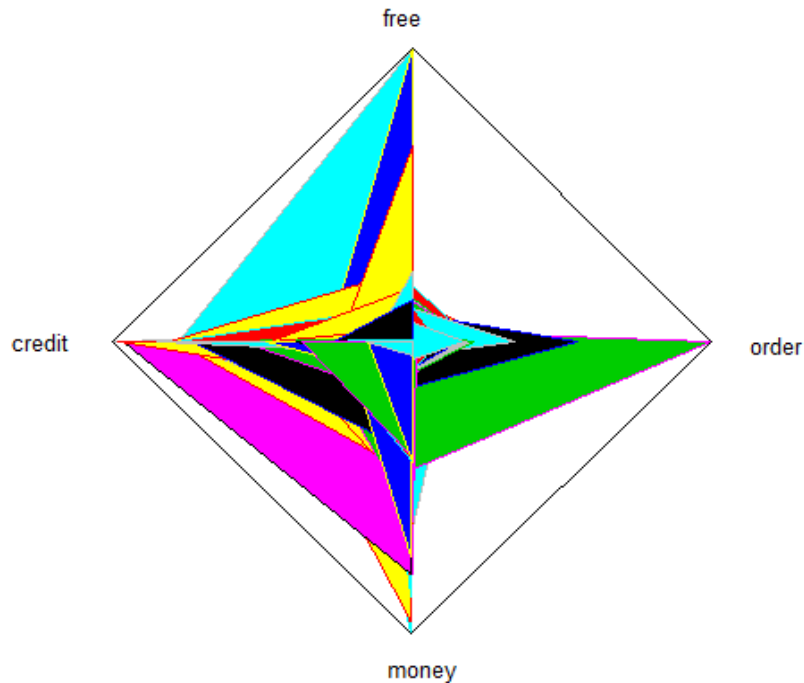
Fig: 'R', package 'Kohonen'  
Intervariate plot of 4 frequent words in Spam

Fig: 'R', package 'SOM'  
Regression of the same four words

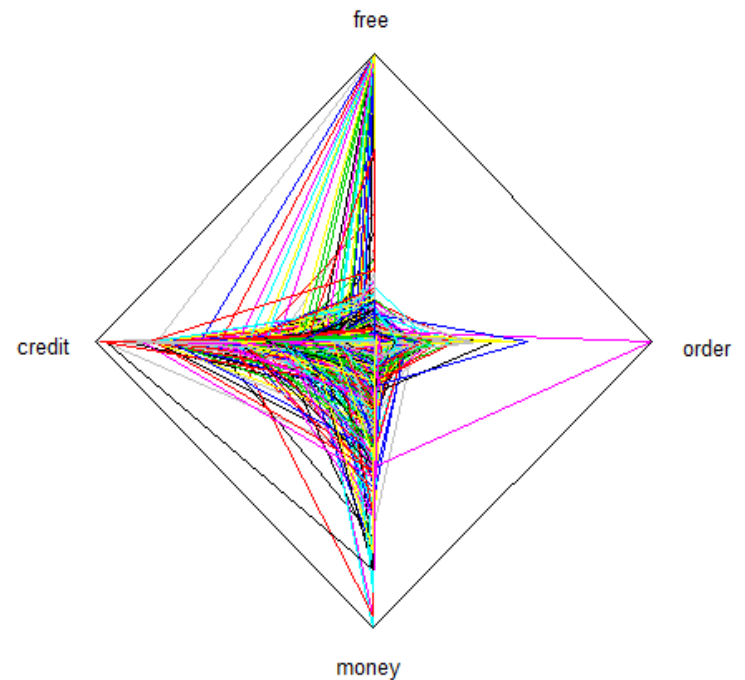
# Algorithm

1. Filter the results
2. Make a polygon with as many sides as the variables.
3. Make the radius of the polygon to be the maximum of the value in the dataset.
4. Draw the grid for the polygon
5. Make segments inside the polygon if the strength of the two variables inside the segment is greater than the specified threshold.
6. Loop Step 5 for every variable against every other variable
7. Color the segments based on the frequency of variable.
8. Color the line segments based on the threshold of each variable pair plotted.

# Spider Plots

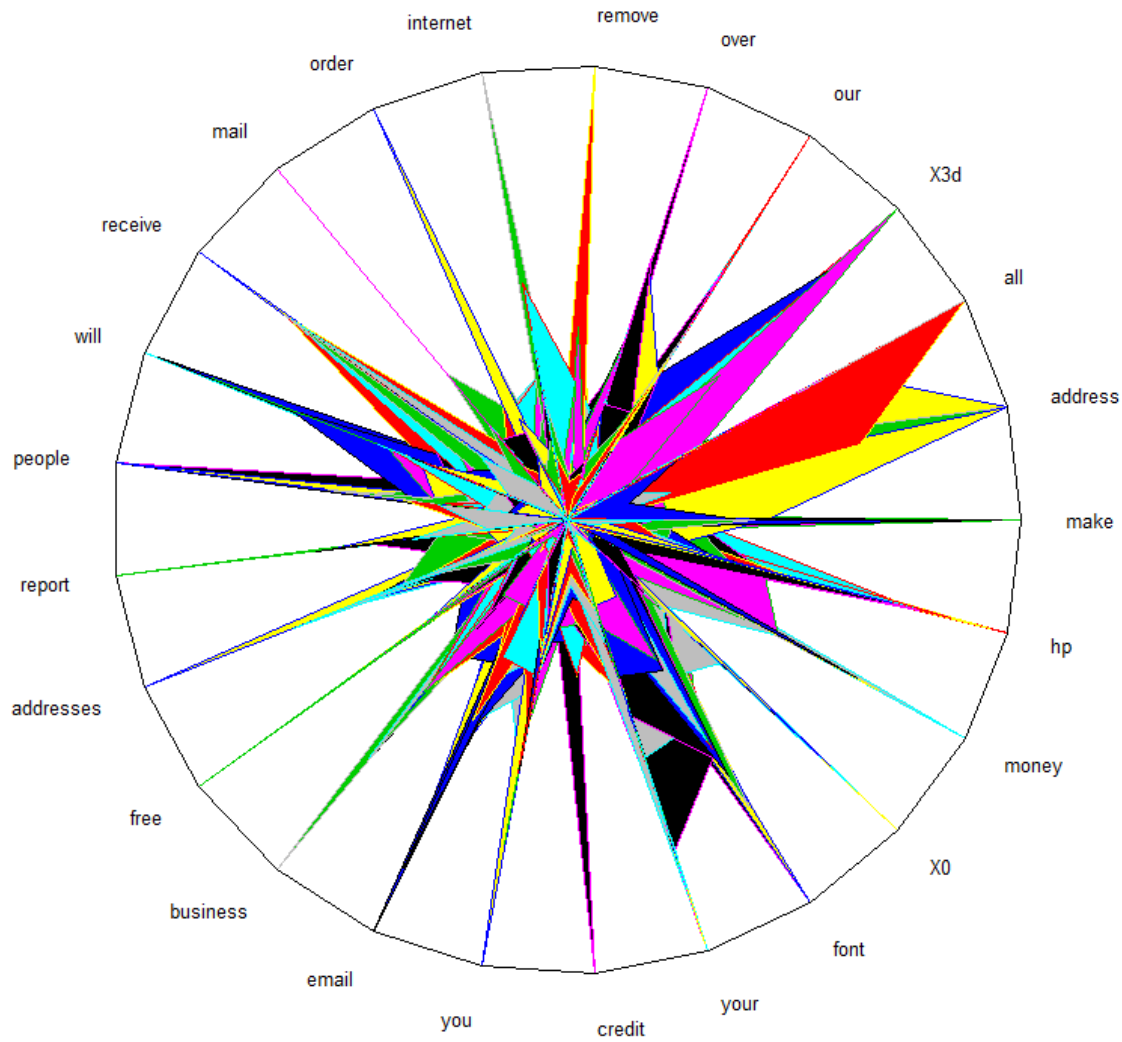


SOM visualization in R using the Algorithm given above. (showing segments i.e inter-variable dependency)



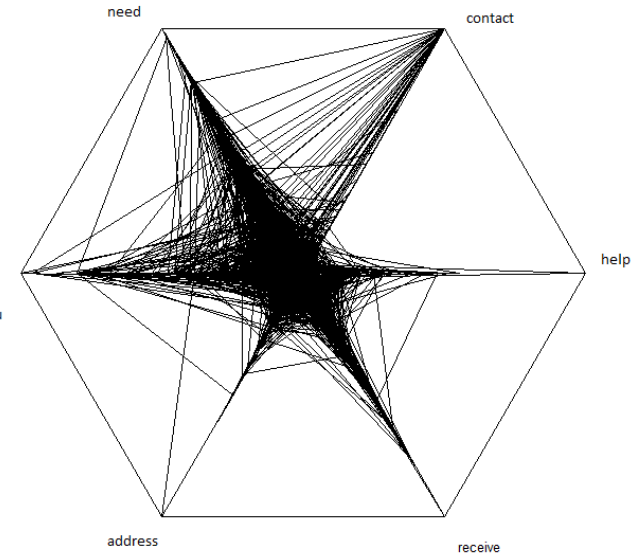
SOM visualization in R using Algorithm given above (showing threads, i.e inter-variable strength)

# Big picture for Big Data



# Conclusion

- Analyze inter-variate relations
- No need to convert Unstructured to Structured Data
- Advantages of Machine Learning and Visualization in single step
- Discover hidden relationships and potentially mining opportunities



# Scope

- Enhance to work with images, sound and videos
- Dynamic representation to show live changes



# References

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