## Label Encoding

Typically, any structured dataset includes multiple columns with combination of numerical as well as categorical variables. A machine can only understand the numbers. It cannot understand the text. That's essentially the case with Machine Learning algorithms too.

We need to convert each text category to numbers in order for the machine to process those using mathematical equations. Label Encoding is a popular encoding technique for handling categorical variables implemented using the scikit-learn library in python. In this technique, each label is assigned a unique integer based on alphabetical ordering.

```
In [115]: from sklearn.preprocessing import LabelEncoder

lb1 = LabelEncoder()
  trainfinal['center_type'] = lb1.fit_transform(trainfinal['center_type'])
  lb2 = LabelEncoder()
  trainfinal['category'] = lb1.fit_transform(trainfinal['category'])
  lb3 = LabelEncoder()
  trainfinal['cuisine'] = lb1.fit_transform(trainfinal['cuisine'])
```

In the above code we have selected text class categorical columns for performing label encoding.

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1     1018704     2     647     56     2     2.0     0     3     135.83     152.29     0     0       2     1196273     3     647     56     2     2.0     0     3     132.92     133.92     0     0       3     1116527     4     647     56     2     2.0     0     3     135.86     134.86     0     0       4     1343872     5     647     56     2     2.0     0     3     146.50     147.50     0     0	1     1018704     2     647     56     2     2.0     0     3     135.83     152.29     0     0       2     1196273     3     647     56     2     2.0     0     3     132.92     133.92     0     0       3     1116527     4     647     56     2     2.0     0     3     135.86     134.86     0     0	1     1018704     2     647     56     2     2.0     0     3     135.83     152.29     0       2     1196273     3     647     56     2     2.0     0     3     132.92     133.92     0       3     1116527     4     647     56     2     2.0     0     3     135.86     134.86     0       4     1343872     5     647     56     2     2.0     0     3     146.50     147.50     0	0	1379560	- 4	647	56	2	2.0	0	3	136.83	152.29		0	
2     1196273     3     647     56     2     2.0     0     3     132.92     133.92     0     0       3     1116527     4     647     56     2     2.0     0     3     135.86     134.86     0     0       4     1343872     5     647     56     2     2.0     0     3     146.50     147.50     0     0	2     1196273     3     647     56     2     2.0     0     3     132.92     133.92     0     0       3     1116527     4     647     56     2     2.0     0     3     135.86     134.86     0     0       4     1343872     5     647     56     2     2.0     0     3     146.50     147.50     0     0	2     1196273     3     647     56     2     2.0     0     3     132.92     133.92     0       3     1116527     4     647     56     2     2.0     0     3     135.85     134.86     0       4     1343872     5     647     56     2     2.0     0     3     146.50     147.50     0		13/9300		047	30	-	2.0			130.03				
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4 1343872 5 647 56 2 2.0 0 3 146.50 147.50 0 0	4 1343872 5 647 56 2 2.0 0 3 146.50 147.50 0 0	4 1343872 5 647 56 2 2.0 0 3 146.50 147.50 0	2	1196273	3	647	56	2	2.0	0	3	132.92	133.92	0	0	
			3	1116527	4	647	56	2	2.0	0	3	135.86	134.86	0	0	
		4	4	1343872	5	647	56	2	2.0	0	3	146.50	147.50	0	0	
			4													

In [117]: trainfinal.shape

Out[117]: (456548, 13)