

Convert Categorical Variables to Numeric in R

Authored by
Mohammed loot

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The ability to effectively manipulate data types is fundamental when working in [R](#). Specifically, converting a [categorical variable](#) (often stored as a [factor](#)) into a numerical format is a common necessity for statistical analysis and machine learning workflows. When [categorical variables](#) are converted to numeric, R assigns an integer based on the factor level ordering.

Understanding these conversion techniques ensures your data is correctly prepared for modeling. We will explore three robust methods for this transformation within an [R data frame](#), moving from targeted conversion to bulk processing.

Method 1: Converting a Single Categorical Variable

This approach uses the built-in `unclass()` function, which is efficient for targeting a specific column within your data structure. This method is ideal when you only need to process one [categorical variable](#) at a time.

```
df$var1 <- unclass(df$var1)
```

Method 2: Converting Multiple Specific Variables

If your analysis requires transforming several, but not all, categorical columns, you can leverage the power of `sapply()` in conjunction with `unclass()`. This allows for efficient, vectorized operations across a defined subset of variables.

```
df <- sapply(df, unclass)
```

Method 3: Transforming All Factor Variables Automatically

For large datasets where manually listing every [factor](#) variable is impractical, [R](#) provides a highly efficient approach. By combining `sapply(df, is.factor)` with `data.matrix()`, we can identify and convert all columns currently stored as factors simultaneously.

```
df <- data.matrix(df)
```

Setting Up the Example Data Frame

To fully demonstrate these three techniques, we will initialize a simple [data frame](#) containing a mix of factor variables and numeric data. In R, it is common practice to explicitly define columns as factors when they represent categories or groups. Using the correct data type is crucial before attempting conversion.

Our example data includes fields for `team`, `conf` (conference), `win` (a binary outcome), and `points` (a standard numeric score). Notice how we use the `as.factor()` function during creation to ensure the variables are properly typed as factors initially.

#create data frame with some categorical variables

```
df <- data.frame(team=as.factor(c('A', 'B', 'C', 'D')),  
conf=as.factor(c('AL', 'AL', 'NL', 'NL')),  
win=as.factor(c('Yes', 'No', 'No', 'Yes')),  
points=c(122, 98, 106, 115))
```

#view data frame

df

team conf win points

1 A AL Yes 122

2 B AL No 98

3 C NL No 106

4 D NL Yes 115

Applying Method 1: Converting a Single Variable

When dealing with a single [factor](#) column, the most straightforward conversion method involves using `unclass()`. The `unclass()` function removes the factor class structure, revealing the underlying integer codes that R uses internally to represent the levels.

In this example, we apply the function specifically to the `team` column. Since 'A' was the first level encountered (assuming alphabetical order), it is converted to 1, 'B' to 2, and so forth. This sequential mapping is critical to remember when interpreting the resulting numerical data, as it dictates the ordinal relationship.

#convert 'team' variable to numeric

```
df$team <- unclass(df$team)
```

#view updated data frame

df

team conf win points

1 1 AL Yes 122

2 2 AL No 98

3 3 NL No 106

4 4 NL Yes 115

Observe how the values in the `team` column have been successfully transformed from character levels (A, B, C, D) into their corresponding integer codes (1, 2, 3, 4). This transformation prepares the variable for use in numerical algorithms.

Applying Method 2: Converting Specific Multiple Variables

To handle several columns simultaneously without iterating manually, we employ the `sapply()` function. This function applies a specified function (in our case, `unclass`) across a list or vector of elements, returning a vector or matrix of the results.

Here, we target both the `team` and `win` columns for conversion. The key advantage of using `sapply` is that it processes the columns efficiently and allows us to assign the results back to the original [data frame](#) subset in a single, clean line of code.

```
#convert 'team' and 'win' variables to numeric
```

```
df <- sapply(df, unclass)
```

```
#view updated data frame
```

```
df
```

```
team conf win points
```

```
1 1 AL 2 122
```

```
2 2 AL 1 98
```

```
3 3 NL 1 106
```

```
4 4 NL 2 115
```

We can confirm that both the `team` and `win` variables now contain numeric values. Note that for the `win` column, the factor levels ('No', 'Yes') have been assigned integer codes (1 and 2, respectively), demonstrating consistent ordinal encoding across multiple selected columns.

Applying Method 3: Converting All Categorical Variables

This method is highly valuable for data cleaning pipelines where you need to standardize all factor variables automatically, regardless of their column names. We use `sapply(df, is.factor)` to create a logical vector identifying all factor columns within the [data frame](#).

The identified factor columns are then transformed using `data.matrix()`. Unlike `unclass()`, `data.matrix()` is specifically designed for coercing factor columns into a matrix structure, which inherently converts factors to their numerical codes while preserving the overall structure for reassignment back into the data frame.

```
#convert all categorical variables to numeric
```

```
df <- data.matrix(df)
```

```
#view updated data frame
```

```
df
```

```
team conf win points
```

```
1 1 1 2 122
```

```
2 2 1 1 98
```

```
3 3 2 1 106
```

```
4 4 2 2 115
```

Upon reviewing the output, all three [factor](#) columns--team, conf, and win--have been successfully converted to numerical representations (1s and 2s), leaving the already numeric points column untouched. This provides a clean, holistic solution for data preparation.

Understanding Encoding and Potential Pitfalls

While converting factors to numeric codes is straightforward, it is essential to understand the underlying encoding mechanism. When [R](#) converts a [factor](#), it uses arbitrary integer values (1, 2, 3...) based on the alphabetical order of the levels, unless the levels were explicitly defined otherwise upon factor creation. This process is effectively **ordinal encoding**.

The primary concern with direct numerical conversion is that it imposes an artificial order and magnitude onto the data. For example, if your conf variable has levels "AL" (coded as 1) and "NL" (coded as 2), statistical models will interpret 2 as being numerically greater than 1. This assumption of linearity is often incorrect for nominal [categorical variables](#), where there is no inherent ranking.

If the categorical data is purely nominal (e.g., colors, names, geographical regions), a more appropriate technique for modeling is **One-Hot Encoding** (or creating dummy variables), which avoids implying an ordinal relationship between categories. If you require dummy variables, specialized functions like `model.matrix()` or packages designed for data preparation are often recommended alternatives.

Summary and Next Steps

We have demonstrated three reliable methods for coercing categorical [factor](#) variables into numeric integers in R. The choice between using `unclass()` for single variables, `sapply()` for specific subsets, or the `data.matrix()` approach for bulk conversion depends entirely on the scale and complexity of your data manipulation needs.

Always verify the resulting data types and the mapping of levels to ensure the conversion aligns with your analytical goals, especially regarding the imposed ordinality of the encoded values.

Additional Resources

The following tutorials explain how to perform other common conversions in R: