

Learning to Create 3D Plots in R: A Step-by-Step Guide

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In the field of data science and statistical analysis, moving beyond two dimensions is often critical for revealing true data structures. When analyzing complex phenomena involving three variables, the generation of a [3D plot](#) provides an unparalleled method for visualizing the relationship between the X, Y, and Z axes simultaneously. This spatial representation is essential when the Z value is a function of both X and Y inputs.

Within the [R programming language](#), the most efficient and fundamental tool for constructing a three-dimensional surface plot is the standard graphics package function: **`persp()`**. This function is specifically engineered to draw perspective plots of surfaces defined over a planar grid. While specialized packages offer interactive features, mastering **`persp()`** is foundational for generating static, publication-quality graphics in R.

Understanding the input structure is key to successful implementation. The **`persp()`** function requires three primary arguments to accurately define the geometry of the surface it intends to render:

`persp(x, y, z)`

Here, **`x`** and **`y`** must be [vectors](#) defining the coordinates across the plane, while **`z`** must be a [matrix](#) defining the height or magnitude of the surface at every intersection point derived from the x and y coordinate vectors. The following examples provide a practical, step-by-step guide on how to implement and extensively customize this powerful visualization tool in R.

Example 1: Generating a Basic 3D Surface Plot

To initiate any three-dimensional surface visualization, we must first establish the coordinate domain (X and Y) and subsequently calculate the Z-values, which mathematically represent the height of the surface across that domain. The Z-matrix must contain $Z_{\{i,j\}}$ values corresponding to the function output at every (X_i, Y_j) point combination.

A crucial function for this setup in R is the [`outer\(\)` function](#). This utility function efficiently calculates the Z-matrix by applying a specified mathematical function across all possible combinations of the X and Y vectors, automating the process of defining the surface geometry.

The code snippet below demonstrates the definition of two coordinate vectors, **`x`** and **`y`**, both spanning from -10 to 10. We then define a simple custom function--calculating the square root of the sum of squares--which naturally results in a cone-like surface shape when plotted. The resulting Z-matrix is then passed directly to **`persp()`** to generate the initial, fundamental [3D plot](#).

#define x and y

`x <- -10:10`

y <- -10:10

#define function to create z-values

```
z_values <- function(x, y) {  
  sqrt(x ^ 2 + y ^ 2)  
}
```

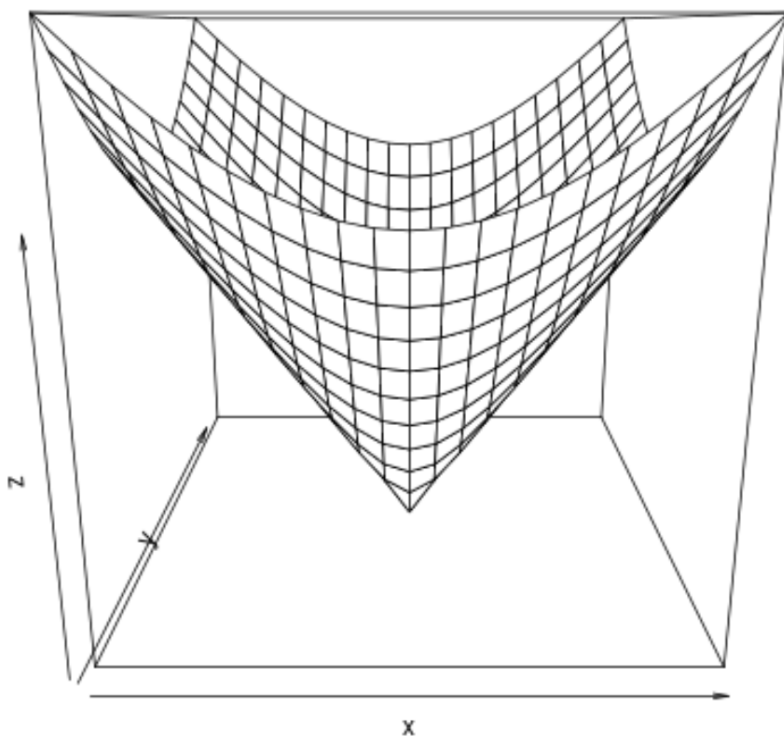
#create z-values

```
z = outer(x, y, z_values)
```

#create 3D plot

```
persp(x, y, z)
```

Executing this command produces the most elementary form of the 3D surface visualization, providing the essential spatial representation required for analysis.



Understanding Arguments: Customizing Visual Aesthetics

While the basic plot successfully renders the surface geometry, effective [data visualization](#) demands enhanced clarity and aesthetic refinement. Visualizations intended for professional reporting or publication must clearly label all dimensions and utilize color and shading to aid interpretation. The [persp\(\) function](#) is highly configurable, offering a comprehensive set of

optional arguments to manage appearance.

Key customization parameters include **xlab**, **ylab**, and **zlab**, which allow the user to apply descriptive textual labels to the corresponding axes, eliminating ambiguity regarding the variables being plotted. Additionally, the **main** argument sets the overall title for the graphic, providing necessary context. Beyond labeling, aesthetic parameters like **col** (surface color) and **shade** (lighting intensity) significantly impact the graphic's quality.

The **shade** argument, in particular, is crucial for rendering 3D plots realistically. By adjusting the intensity of the light source, the surface texture and contours become more pronounced, effectively enhancing the perception of depth and curvature--a critical component in interpreting complex surfaces. The following example applies these parameters to transform the basic wireframe into a visually informative graphic.

```
#define x and y
```

```
x <- -10:10
```

```
y <- -10:10
```

```
#define function to create z-values
```

```
z_values <- function(x, y) {
```

```
  sqrt(x ^ 2 + y ^ 2)
```

```
}
```

```
#create z-values
```

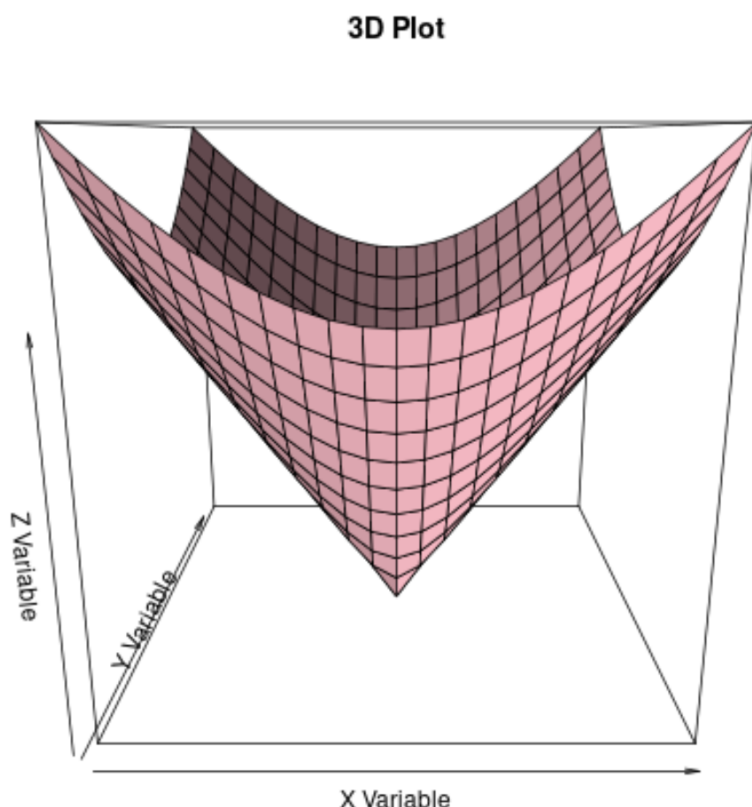
```
z = outer(x, y, z_values)
```

```
#create 3D plot
```

```
persp(x, y, z, xlab='X Variable', ylab='Y Variable', zlab='Z Variable',
```

```
main='3D Plot', col='pink', shade=.4)
```

These detailed aesthetic improvements ensure that the graphic is not only functional but also immediately understandable to the intended audience, moving the visualization toward publication quality.



Controlling Viewer Perspective: Rotation with Theta and Phi

A fundamental challenge in visualizing three dimensions on a two-dimensional screen is the static nature of the image. The default viewing angle provided by **persp()** might inadvertently obscure key structural characteristics or critical data points on the surface. To mitigate this issue, the function allows the user to dynamically control the viewing perspective through two critical rotational arguments: **theta** and **phi**.

The **theta** argument controls the horizontal rotation, defining the [azimuthal direction](#) of the viewpoint. This is equivalent to rotating the plot around the Z-axis. Conversely, the **phi** argument controls the vertical rotation, known as the [colatitude](#), which specifies the angle of the viewpoint relative to the x-y plane. Adjusting these parameters is essential for exploratory analysis, as it allows the user to iteratively search for the optimal angle that most effectively highlights the nuances of the three-dimensional relationship.

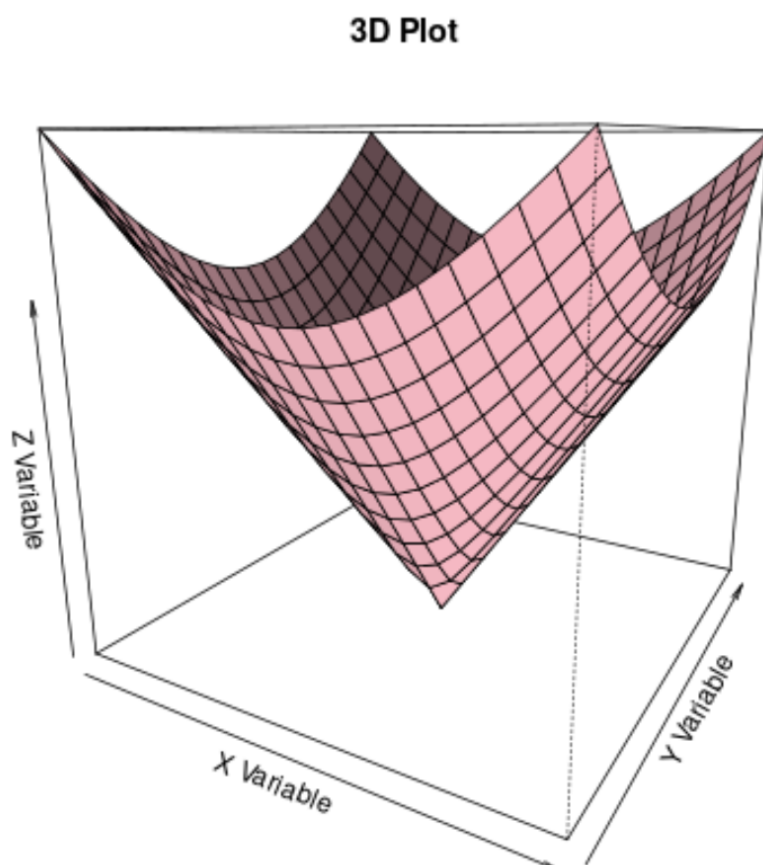
In the following script, we maintain the comprehensive aesthetic customizations from the previous example but introduce specific values for rotation: `theta = 30` and `phi = 15`. This intentional manipulation shifts the viewpoint, often resulting in a more informative and analytically accessible representation of the surface structure.

```
#define x and y
x <- -10:10
y <- -10:10

#define function to create z-values
z_values <- function(x, y) {
  sqrt(x ^ 2 + y ^ 2)
}

#create z-values
z = outer(x, y, z_values)

#create 3D plot
persp(x, y, z, xlab='X Variable', ylab='Y Variable', zlab='Z Variable',
main='3D Plot', col='pink', shade=.4, theta = 30, phi = 15)
```



Achieving Quantitative Clarity with Detailed Tick Marks

While the previous examples focused on visual structure and aesthetic appeal, a visualization is incomplete if it lacks specific numerical anchors for quantitative interpretation. The default settings of many plotting functions, including **persp()**, prioritize a clean visual appearance, often at the expense of precise numerical context. This means the internal axes might lack specific numerical indicators (tick marks and labels).

To ensure quantitative readability, the **ticktype** argument within the [persp\(\) function](#) must be utilized. By default, **ticktype** is set to 'simple', which often results in sparse labeling. However, setting this argument to 'detailed' forces the systematic display of numerical tick marks and corresponding labels along all three axes (X, Y, and Z).

This simple but vital change dramatically improves the plot's utility for rigorous analysis, allowing viewers to quickly associate specific points on the surface with their corresponding coordinate values. This final example integrates the advanced rotational settings (theta and phi) with the essential detailed tick marks, resulting in a comprehensive and fully labeled visualization suitable for professional reporting in the [R programming language](#).

#define x and y

```
x <- -10:10
```

```
y <- -10:10
```

#define function to create z-values

```
z_values <- function(x, y) {
```

```
  sqrt(x ^ 2 + y ^ 2)
```

```
}
```

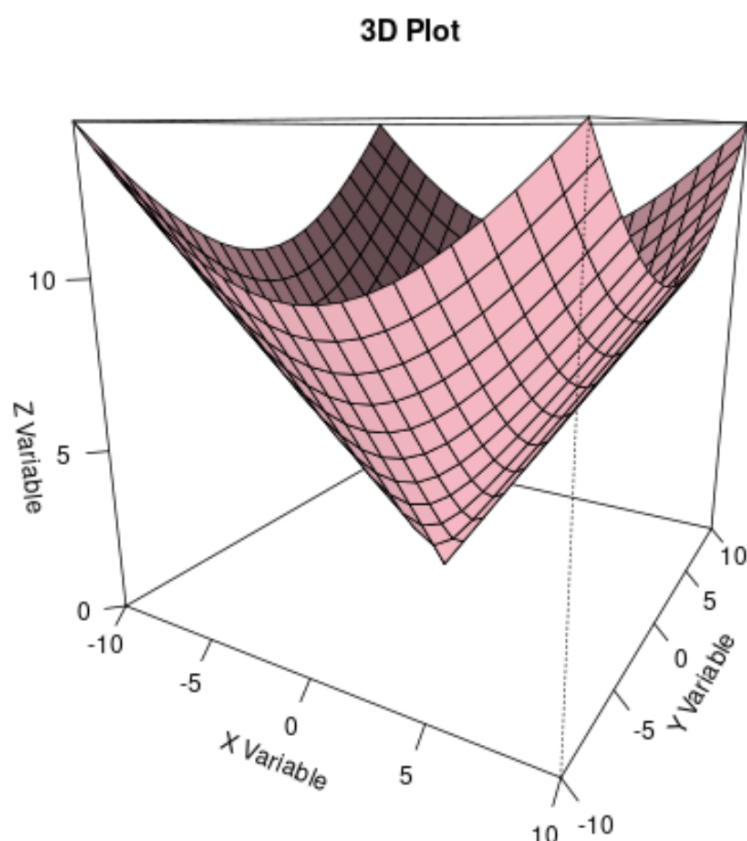
#create z-values

```
z = outer(x, y, z_values)
```

#create 3D plot

```
persp(x, y, z, xlab='X Variable', ylab='Y Variable', zlab='Z Variable',
```

```
main='3D Plot', col='pink', shade=.4, theta = 30, phi = 15, ticktype='detailed')
```



Conclusion: Summary and Next Steps in 3D Visualization

The **`persp()`** function, residing within the base graphics system of R, provides a fundamentally robust and highly flexible method for generating static 3D surface plots. By mastering its core arguments--specifically those that define the geometry (x, y, z), control the viewer's orientation (theta, phi), and enhance readability (xlab, zlab, ticktype)--users can effectively transform raw three-dimensional coordinate data into insightful and aesthetically pleasing visualizations.

For users whose visualization requirements extend beyond static images, such as requiring dynamic rotation, zooming, or web-embedded graphics, alternative packages in R should be explored. Specialized libraries like **`rgl`** (for high-performance OpenGL graphics) or **`plotly`** (for interactive, web-based plots) offer expanded capabilities. However, for quick prototyping, reproducible scripts, and effective surface mapping within the standard R environment, **`persp()`** remains an indispensable and efficient tool in the R data scientist's toolkit.

To further advance your skills in data visualization within the R ecosystem, consider exploring methods for generating other complex chart types:

How to create complex scatter plots incorporating grouping and trends.

Generating heatmaps for advanced correlation and density analysis.

Designing effective bar charts and histograms for categorical and distributional data.