Let X and Y be two independent normally distributed random variables with expected value 0 and variance 1. Find their joint PDF. Plot its level curves.

Since X and Y are independent, their joint PDF is simply the product of their individual PDFs:

$$pdf_{X,Y}(x,y) = pdf_{X}(x) \cdot pdf_{Y}(y) = \left(\frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{x^{2}}{2}}\right) \cdot \left(\frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{y^{2}}{2}}\right) = \frac{1}{2\pi} \cdot e^{-\frac{x^{2} + y^{2}}{2}}$$

This is the joint PDF of a bivariate normal distribution with mean (expectation) $\mu = (0, 0)$ and covariance matrix: $\Sigma = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$.

We can plot level curves and the shape of the joint *PDF* using this code snippet:

```
1
   import numpy as np
   import matplotlib.pyplot as plt
   from scipy.stats import multivariate normal
3
4
5
   mean = [0, 0]
   cov_matrix = [[1, 0], [0, 1]]
6
   data = multivariate_normal(mean, cov_matrix)
8
9
   x, y = np.mgrid[-2:2:.01, -2:2:.01]
   xy_position = np.dstack((x, y))
10
11
   plt.contourf(x, y, data.pdf(xy_position))
12
13
   plt.colorbar()
14
   plt.figure().add_subplot(projection='3d')
15
   plt.contourf(x, y, data.pdf(xy_position), 200);
```

The resulting plots themselves will have the following look:

