Machine Learning

HW#1

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(a) Let the event that I want to go up be The Let the event that I want to go down be Me Let the event that the lift is going down be I Ele Now, the required probability is equivalent to: P((1Me 1 TELE) V (VMe 1 VELE)) Let k+1 be the floor So, 0 = k & h Now, two cases: i) k \le \frac{7}{2} I-P (none in ana) p (TELe) = P (TELe 1 2 tele
inshaded + P (TELe 1 none in ) Shaded ama X P ( nome ? P(VEU) = 1 - P(TEU)  $= 1 - \left(\frac{1}{2} \times \left(1 - \left(\frac{n-2k}{n}\right)^m\right)\right)$ 

$$P(VELe) = P(VELe) = P(VELe) = \frac{1}{4} \frac{1}{1} \frac{1}{1$$

By Symmetry:
$$P((1 \text{ Me } 1 \text{ Tele}) \vee (1 \text{ Me } 1 \text{ Vele}))$$

$$= 2 \sum_{k=0}^{n_{1}} \left[ \frac{n-k}{n} \times \left( \frac{1}{2} \times (1 - \left( \frac{n-2k}{n} \right)^{m} \right) \times \frac{1}{n+1} \right]$$

$$+ 2 \sum_{k=0}^{n_{1}} \left[ \frac{n}{n} \times \left( 1 - \left( \frac{1}{2} \times \left( 1 - \left( \frac{n-2k}{n} \right)^{m} \right) \right) \times \frac{1}{n+1} \right]$$

$$= 2 \sum_{k=0}^{n_{1}} \left[ \frac{n-k}{n} \times \frac{1}{2} \times \left( 1 - \left( \frac{n-2k}{n} \right)^{m} \right) \times \frac{1}{n+1} \right]$$

$$+ \frac{k}{n} \times \left( 2 - \left( \frac{1}{2} \times \left( 1 - \left( \frac{n-2k}{n} \right)^{m} \right) \right) \times \frac{1}{n+1}$$

$$= 2 \sum_{k=0}^{n_{1}} \left[ \frac{(n-k)}{2n} \left( 1 - \left( \frac{n-2k}{n} \right)^{m} \right) + k \left( 2 - 1 + \left( \frac{n-2k}{n} \right)^{m} \right) \right]$$

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Now, the two elevators can be on any of the floor and traveling strictly up or down. However, for the next elevator to come to me must be traveling downwards in order for me to use It. For example, it an elevator is currently 4th floor and traveling downwards, it needs to reach 1st floor, then go up to 9th floor and then come down to me on the 7-72 Ploor. We can simplify this problem into something analogous to a the face of a clock as follows:

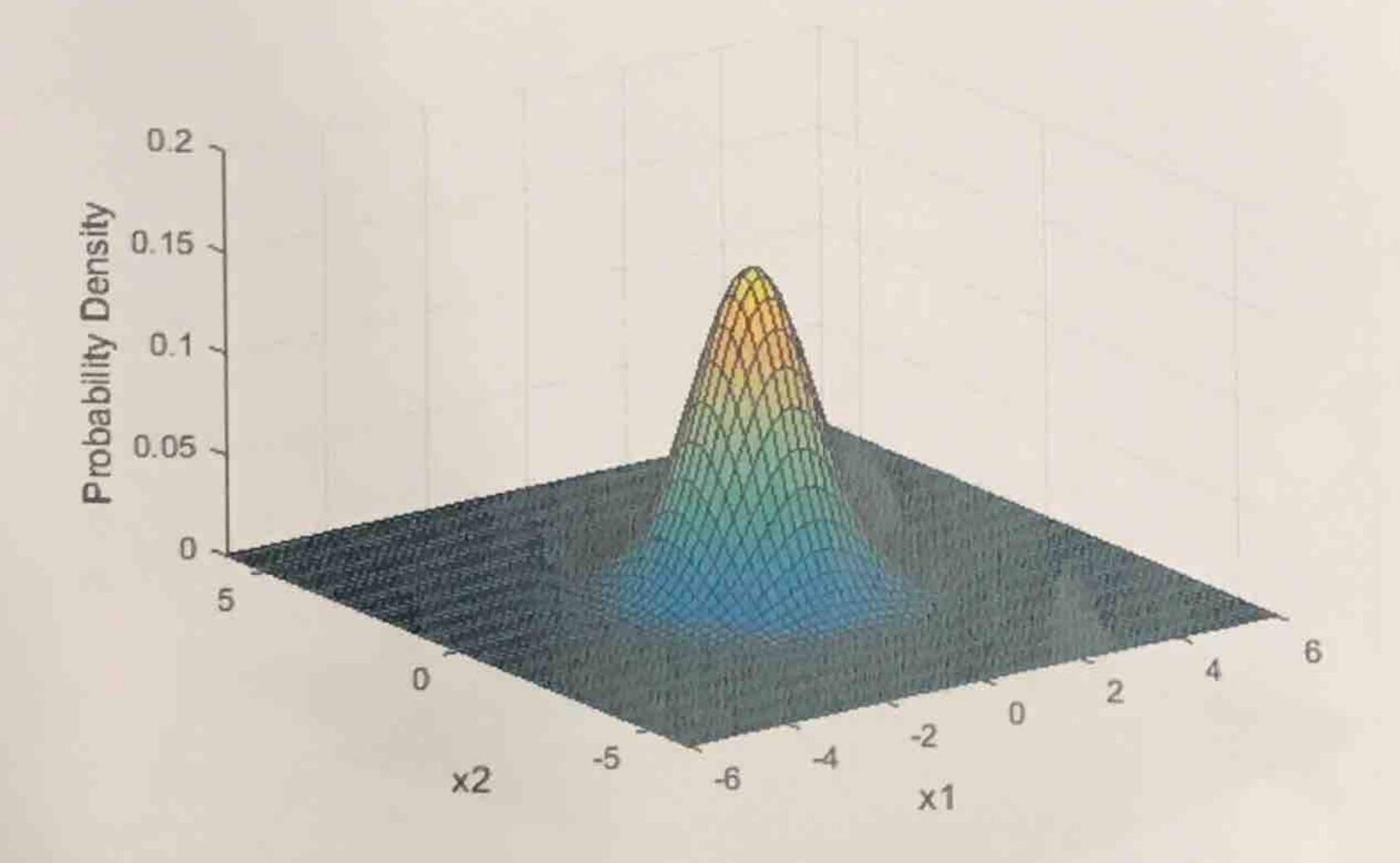
Assuming we travel only andiclockwise on the clock date, we can easily see the above example fitting nicely on this clock face.

Now, if we had only one elevator, and this elevators is equally likely to be on any floor then by law of total expectation, it is expected to take 8 minutes to reach me since there are 16 floors and each floor takes I minute - On the clock face lethis means the elevator is expected to be on Bud floor and travelling up. Now, thee secondal elevator can be anywhere between me and the first elevator, and the expeted time is thus 4 minutes. In terms of expetation of joint probability Let x be first elevator 11 7 11 Second elevation Now, ELXYJ = EBJ ECYJ  $=\frac{1}{2}\times\frac{16}{2}$ = 4 min [ms]

```
2.a.I.i
mu = [0 0];
Sigma = [1 0; 0 1];
x1 = -6:.2:6; x2 = -6:.2:6;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));
surf(x1,x2,F);
caxis([min(F(:))-.5*range(F(:)), max(F(:))]);
```

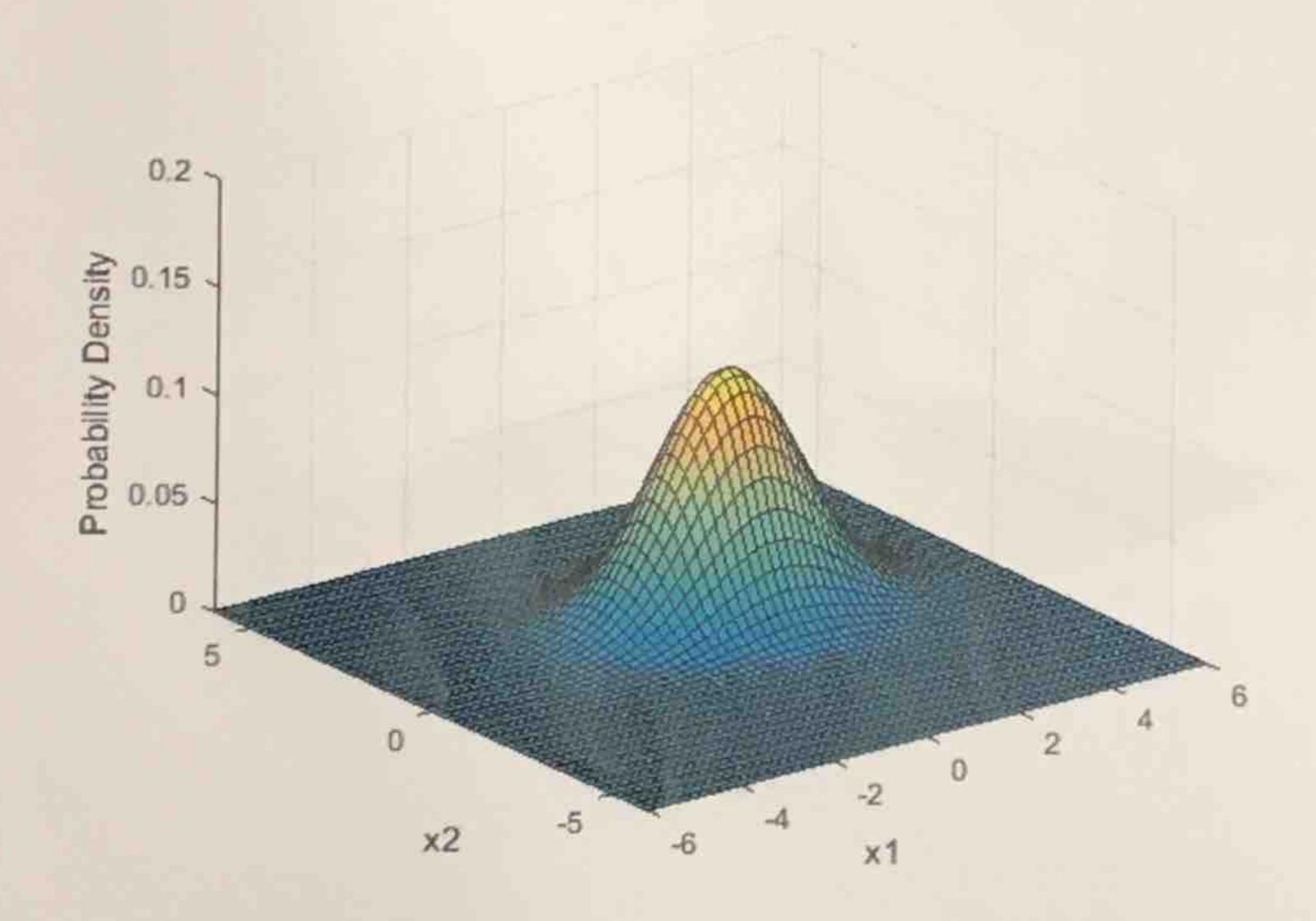
xlabel('x1'); ylabel('x2'); zlabel('Probability Density');

axis([-6 6 -6 6 0 .2])



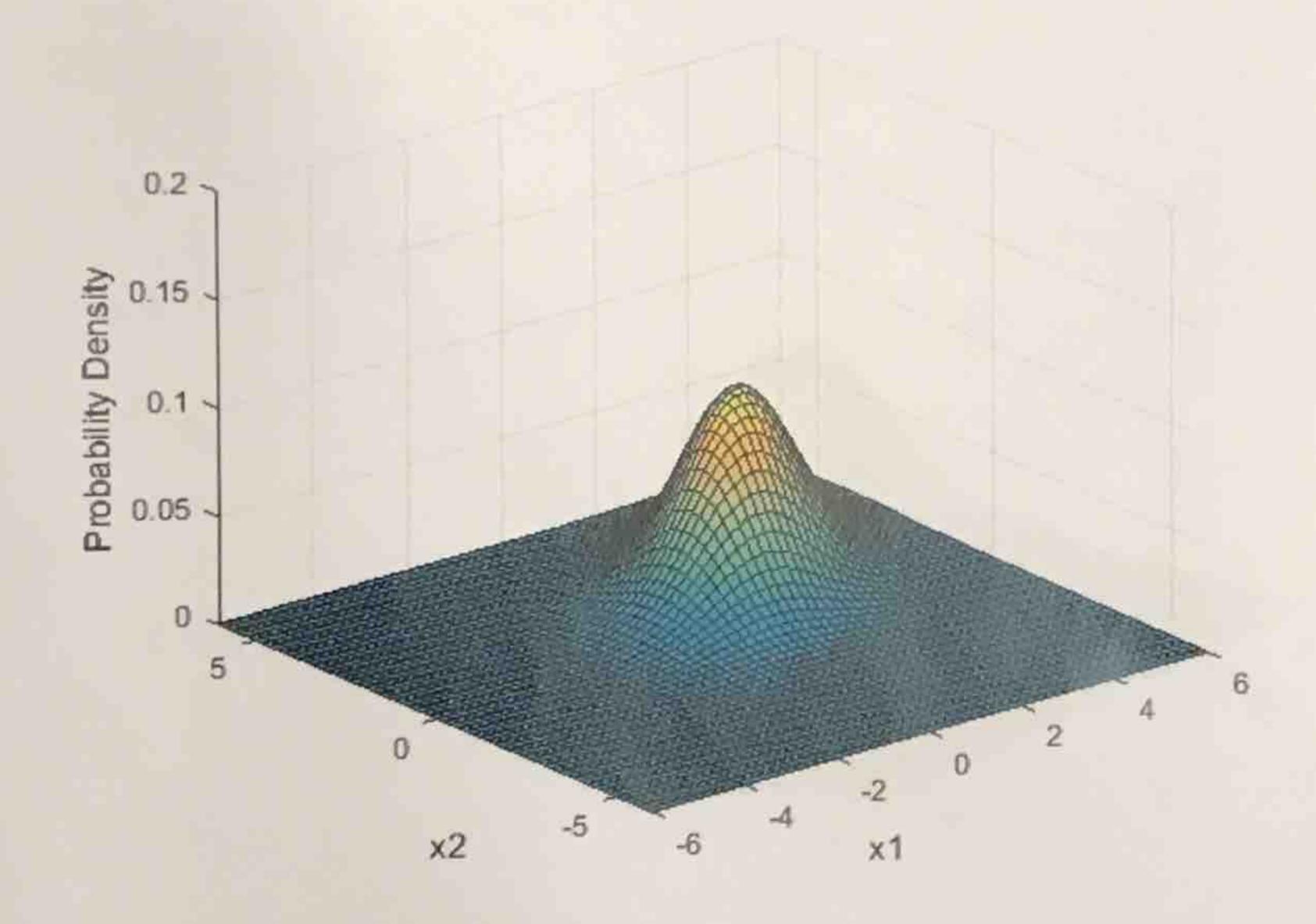
## 2.a.I.ii

```
mu = [1 1];
Sigma = [2 0; 0 1];
x1 = -6:.2:6; x2 = -6:.2:6;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));
surf(x1,x2,F);
caxis([min(F(:))-.5*range(F(:)),max(F(:))]);
axis([-6 6 -6 6 0 .2])
xlabel('x1'); ylabel('x2'); zlabel('Probability Density');
```



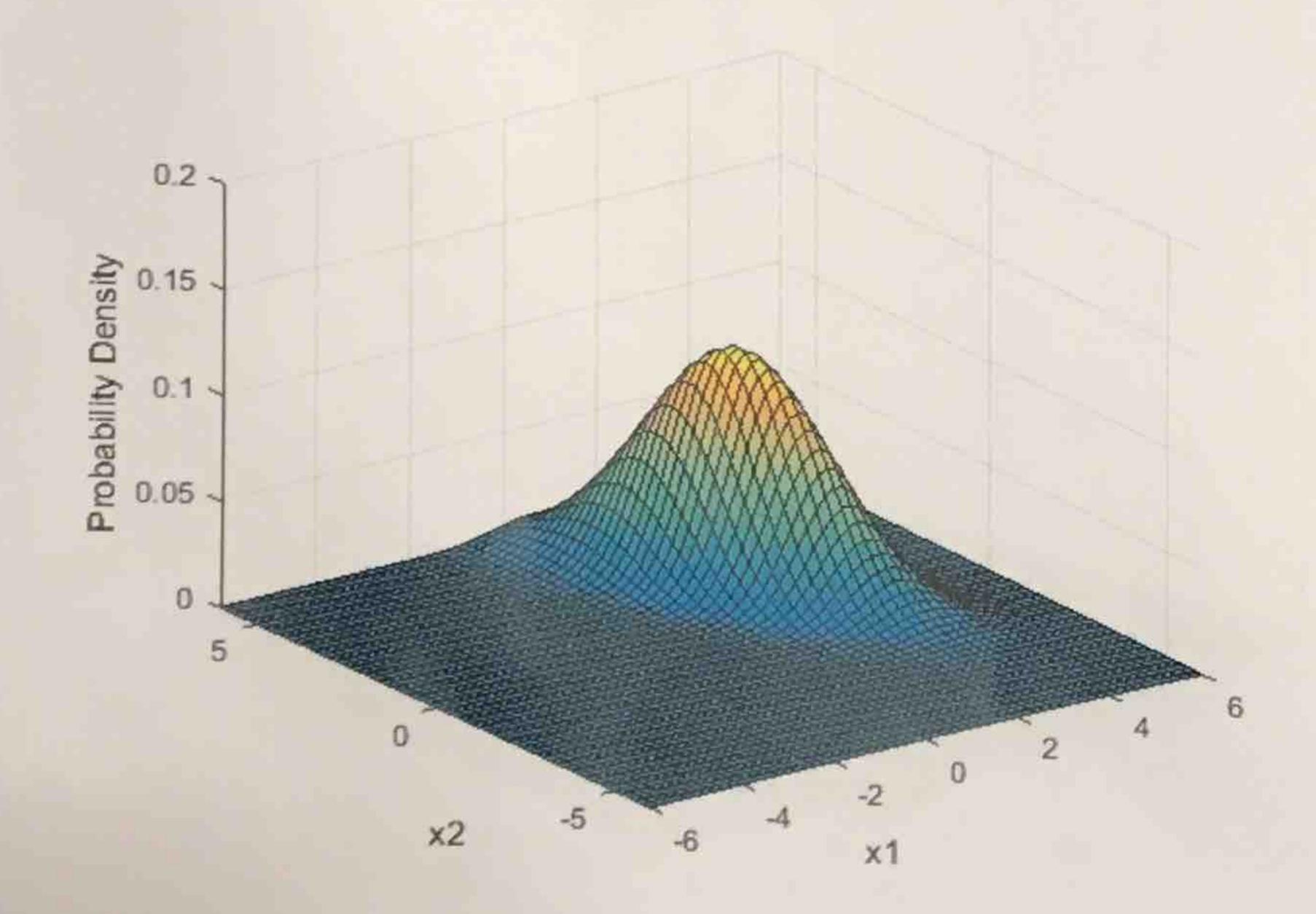
```
2.a.I.iii
```

```
mu = [2 2];
Sigma = [2 0.95; 0.95 2];
x1 = -6:.2:6; x2 = -6:.2:6;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));
surf(x1,x2,F);
caxis([min(F(:))-.5*range(F(:)),max(F(:))]);
axis([-6 6 -6 6 0 .2])
xlabel('x1'); ylabel('x2'); zlabel('Probability Density');
```



```
2.a.I.iv

mu = [2 2];
Sigma = [1 -1; -1 3];
x1 = -6:.2:6; x2 = -6:.2:6;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));
surf(x1,x2,F);
caxis([min(F(:))-.5*range(F(:)),max(F(:))]);
axis([-6 6 -6 6 0 .2])
xlabel('x1'); ylabel('x2'); zlabel('Probability Density');
```



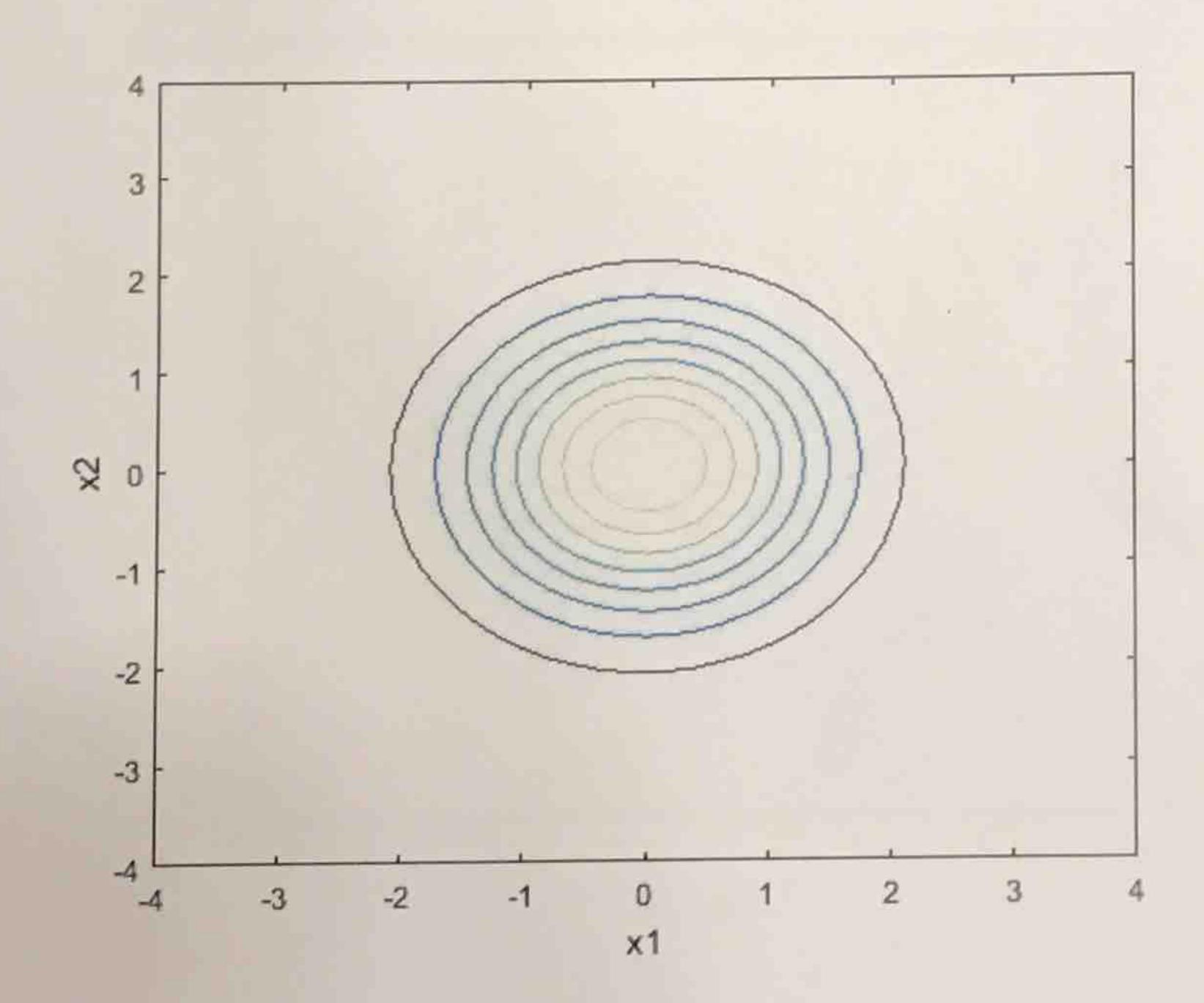
```
2.a.II.i

mu = [0 0];
Sigma = [1 0; 0 1];

max_c = 1/(sqrt(((2*pi)^2) * det(Sigma)))
c_range = linspace(0, max_c, 10)

x1 = -4:.2:4; x2 = -4:.2:4;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));

mvncdf([0 0],[1 1],mu,Sigma);
contour(x1,x2,F,c_range);
xlabel('x1'); ylabel('x2');
```



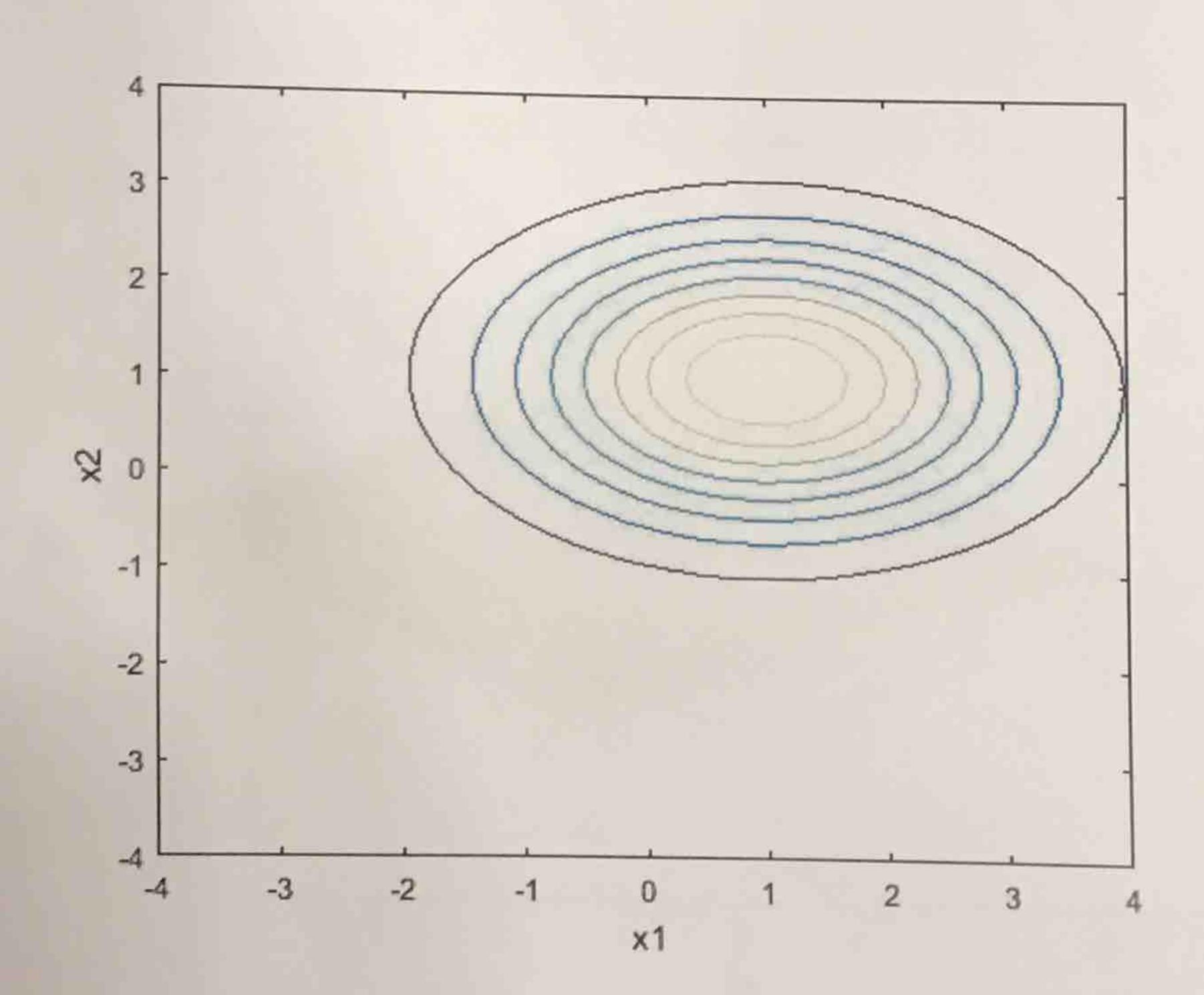
```
2.a.II.ii

mu = [1 1];
Sigma = [2 0; 0 1];

max_c = 1/(sqrt(((2*pi)^2) * det(Sigma)))
c_range = linspace(0, max_c, 10)

x1 = -4:.2:4; x2 = -4:.2:4;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));

mvncdf([0 0],[1 1],mu,Sigma);
contour(x1,x2,F, c_range);
xlabel('x1'); ylabel('x2');
```



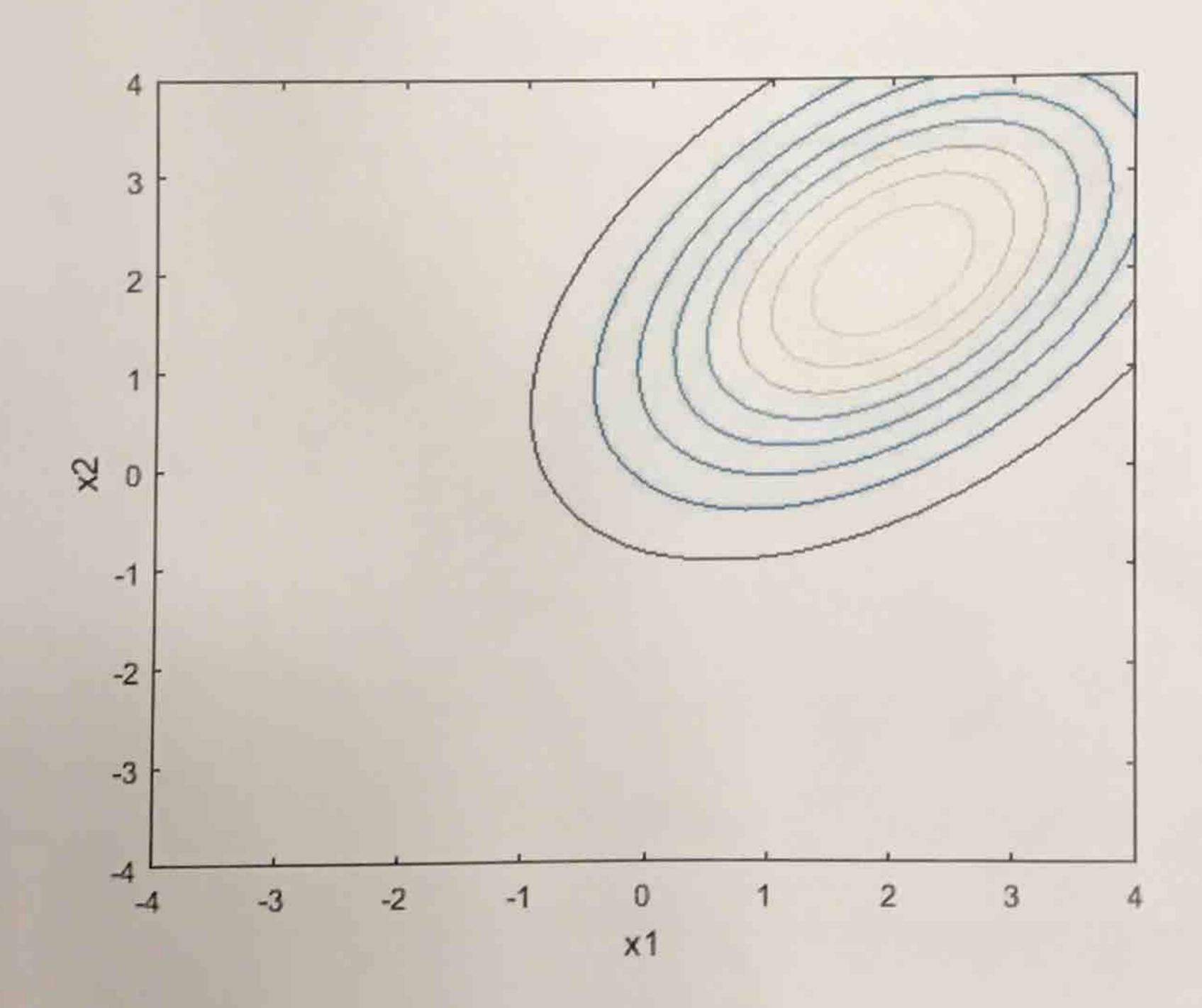
```
2.a.II.iii

mu = [2 2];
Sigma = [2 0.95; 0.95 2];

max_c = 1/(sqrt(((2*pi)^2) * det(Sigma)))
c_range = linspace(0, max_c, 10)

x1 = -4:.2:4; x2 = -4:.2:4;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)],mu,Sigma);
F = reshape(F,length(x2),length(x1));

mvncdf([0 0],[1 1],mu,Sigma);
contour(x1,x2,F,c_range);
xlabel('x1'); ylabel('x2');
```



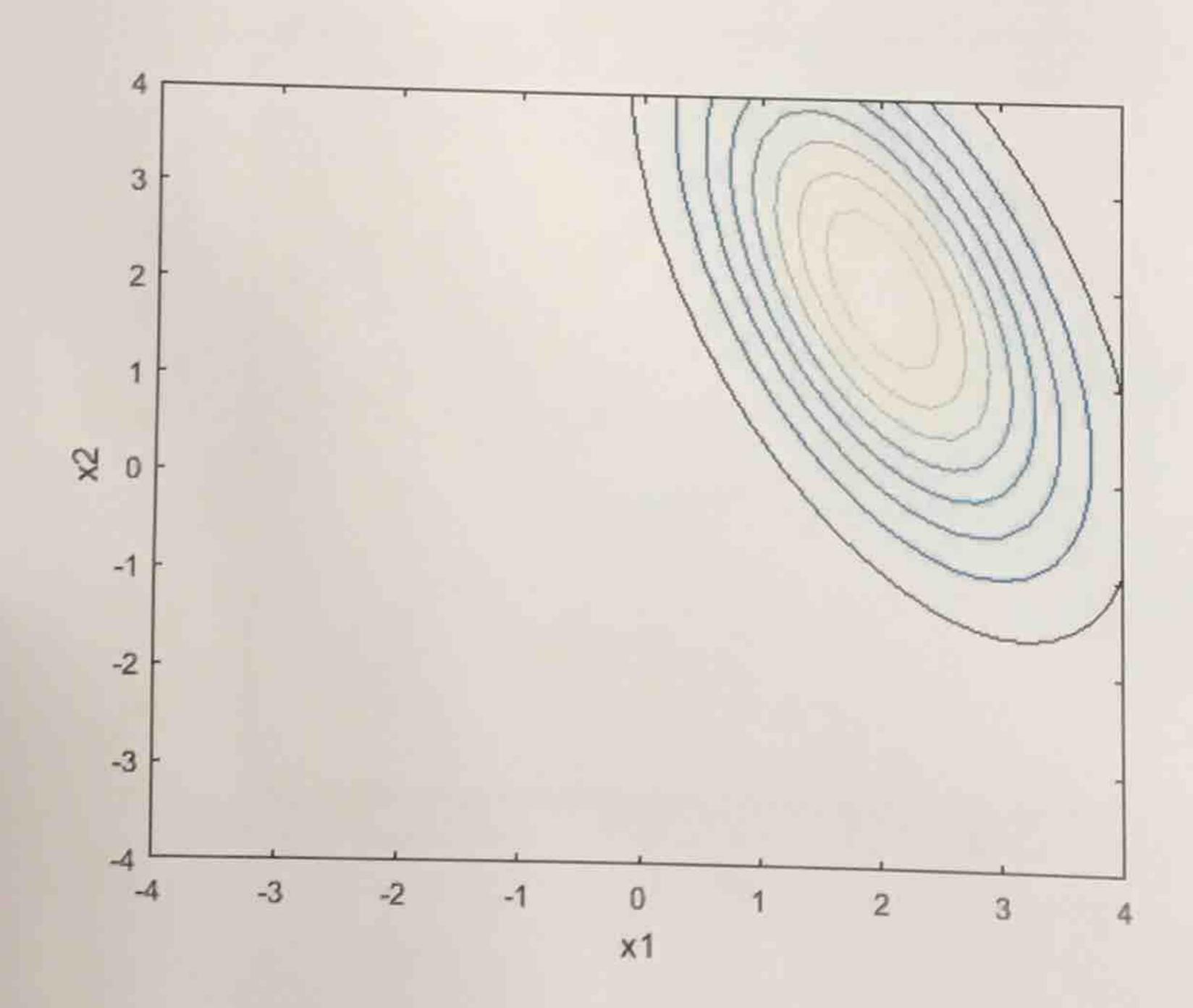
```
2.a.II.iv

mu = [2 2];
Sigma = [1 -1; -1 3];

max_c = 1/(sqrt(((2*pi)^2) * det(Sigma)))
c_range = linspace(0, max_c, 10)

x1 = -4:.2:4; x2 = -4:.2:4;
[X1,X2] = meshgrid(x1,x2);
F = mvnpdf([X1(:) X2(:)], mu, Sigma);
F = reshape(F,length(x2),length(x1));

mvncdf([0 0],[1 1],mu,Sigma);
contour(x1,x2,F,c_range);
xlabel('x1'); ylabel('x2');
```



No, this is not a valle Gaussian. This is because the covaniance matrix of is wrong. This is because by definition, each matrix cell of  $\Phi$ ,  $\forall ij = E[(\kappa_i - \kappa_i)(\kappa_j - \kappa_i)]$ . This means that Si; must be equivalent to Ji as product of KX y is same as yxx, However in the given \$, Jij \$ 5; when i #j. Also, when i=j, (ki-M2) (kj-M3) = (kj-Mg)2, and thus Jij when J=j must be a positive number, Howeve T22 in the given & is -1, which is wrong. Lastly, the determinant of the given \$ 1's negative and we cannot square voot a negative number while calculating the multivariate Gaussian probability doing the function.

multivariate hansstan We can represent the PDF as Px (X; M, D) We will do trese in parts: ) simply odputs X So, 0.5 x Px (x; [0], [0]) multiplies x, by 2 and ontputs Y = [2x, K2]  $y = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ covariance of this Now, we need to find the mean and ECM = E((2.0) [X]) = (2 i) E[A] (Y- Mg) (Y- Mg) (Y- Mg)  $= \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ y_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$ 

Therefore,

$$0.2 \times P_{X}(X; 0, \begin{bmatrix} u & 0 \\ 0 & 1 \end{bmatrix})$$
 $Y = \begin{bmatrix} x_{1} \\ x_{1} + z \end{bmatrix} = \begin{bmatrix} x_{1} \\ x_{2} \end{bmatrix} + \begin{bmatrix} 0 \\ 2 \end{bmatrix}$ 
 $E[Y] = E[X] + E[2]$ 
 $= [X] + [E[2]]$ 
 $= 0 + [0]$ 
 $= [X] + [X]$ 

$$E\left(\left(X-M_{N}\right)\left(R-M_{N}\right)^{T}\right)+E\left(\left(2-M_{2}\right)\left(2-M_{2}\right)^{T}\right)$$

$$+E\left(\left(X-M_{N}\right)\left(2-M_{2}\right)^{T}\right)+E\left(\left(2-M_{2}\right)\left(X-M_{N}\right)^{T}\right)$$

$$=I+\left[0\right]$$

$$=\left[0\right]$$

$$=\left[0\right]$$
So, we have:
$$0.3\times P_{X}\left(X;\left[0\right],\left[0\right]\right)$$
Therefore,
$$P_{Y}=0.5\times P_{X}\left(X;\left[0\right],\left[0\right]\right)$$

$$P_{x} = 0.5 \times P_{x}(x; 0, [0])$$
+  $0.2 \times P_{x}(x; 0, [0])$ 
+  $0.3 \times P_{x}(x; [0])$ 
 $(0.7)$ 

[Ans]