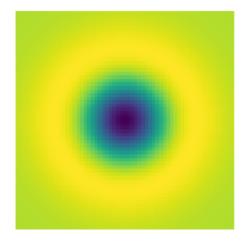
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COURSE CODE: EN2550

```
In [ ]:
         # Importing libraries
         import matplotlib.pyplot as plt
         import cv2 as cv
         import numpy as np
         from matplotlib import cm
In [ ]:
         # question 01
         sigma = 10
         hw = 3*sigma
         X = np.arange(-hw, hw+1, 1)
         Y = np.arange(-hw, hw+1, 1)
         XX, YY = np.meshgrid(X, Y)
         log = 1/(2*np.pi*sigma**2)*(XX**2/(sigma**2)+YY**2/(sigma**2)-2) * \
             np.exp(-(XX**2+YY**2)/(2*sigma**2))
         plt.imshow(log)
         plt.axis("off")
```

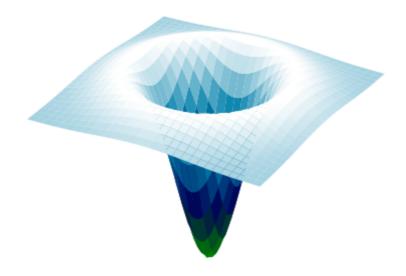
Out[]: (-0.5, 60.5, 60.5, -0.5)



```
#Showing in 3-d
fig = plt.figure(figsize=(18, 18))
ax1 =fig.add_subplot(121,projection='3d')
surf = ax1.plot_surface(XX,YY,log, cmap=cm.ocean,lineWidth=0,antialiased=True)
ax1.axis('off')
```

<ipython-input-6-5dfebaa40661>:4: MatplotlibDeprecationWarning: Case-insensitive pro
perties were deprecated in 3.3 and support will be removed two minor releases later
 surf = ax1.plot_surface(XX,YY,log, cmap=cm.ocean,lineWidth=0,antialiased=True)

Out[]: (-33.0, 33.0, -33.0, 33.0)



```
In []: # question 02
w, h = 71, 71
hw, hh = w//2, h//2

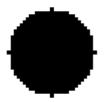
X = np.arange(-hh, hh+1, 1)
Y = np.arange(-hw, hw+1, 1)
XX, YY = np.meshgrid(X, Y)

f = np.ones((h, w), dtype=np.float32)*255

r = w//5
f *= XX**2 + YY**2 > r**2

plt.imshow(f,cmap='gray')
plt.axis('off')
```

Out[]: (-0.5, 70.5, 70.5, -0.5)



```
In [ ]: | S = 11
                         fig, ax = plt.subplots(2, s, figsize=(20, 5))
                         scale_space = np.empty((h, w, s), dtype=np.float32)
                         sigmas = np.arange(5, 16, 1)
                         for i, sigma in enumerate(sigmas):
                                    log_hw = 3*np.max(sigmas)
                                    X = np.arange(-log_hw, log_hw+1, 1)
                                    Y = np.arange(-log_hw, log_hw+1, 1)
                                    XX, YY = np.meshgrid(X, Y)
                                    log = 1/(2*np.pi*sigma**2)*(XX**2/(sigma**2)+YY**2 / (sigma**2)+YY**2 / (sigma**2)*(XX**2/(sigma**2)+YY**2 / (sigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma**2)*(xigma*
                                                                                                                 (sigma**2)-2) * np.exp(-(XX**2+YY**2)/(2*sigma**2))
                                    f_log = cv.filter2D(f, -1, log)
                                    scale_space[:, :, i] = f_log
                                    ax[0, i].imshow(log)
                                    ax[0, i].axis('off')
                                    ax[0, i].set_title(r"$\sigma={}$".format(sigma))
                                    ax[1, i].imshow(f_log,cmap='gray')
                                    ax[1, i].axis('off')
                         indices = np.unravel_index(
                                    np.argmax(scale_space, axis=None), scale_space.shape)
In [ ]:
                         print(indices)
                         print(sigmas[indices[2]])
                       (35, 35, 5)
                       10
                     value of \sigma corresponding to maximum occurance of peak response = \frac{r}{\sqrt{2}} = \frac{14}{\sqrt{2}} = 9.89
                     which is more closure to \sigma = 10
In [ ]:
                         # question 03
                         f3_1=cv.imread(r'images/e5_img1.ppm',cv.IMREAD_GRAYSCALE)
                         assert f3 1 is not None
                         f3_2=cv.imread(r'images/e5_img4.ppm',cv.IMREAD_GRAYSCALE)
                         assert f3_2 is not None
                         sift = cv.SIFT create()
                         keypoints_1, descriptors_1 = sift.detectAndCompute(f3_1,None)
                         keypoints_2, descriptors_2 = sift.detectAndCompute(f3_2,None)
                         #feature matching
```

bf = cv.BFMatcher(cv.NORM_L1, crossCheck=True)

matches = bf.match(descriptors_1,descriptors_2)
matches = sorted(matches, key = lambda x:x.distance)

fig = plt.figure(figsize=(18, 6))

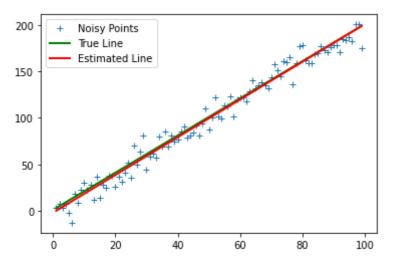
f3 = cv.drawMatches(f3_1, keypoints_1, f3_2, keypoints_2, matches[:50], f3_2, flags=

```
plt.imshow(f3)
plt.axis("off")
plt.show()
```



```
In [ ]:
         # Least square line fitting
         m = 2
         x = np.arange(1, 100, 1)
         np.random.seed(45)
         sigma = 10
         n = sigma*np.random.randn(len(x))
         o = np.zeros(x.shape)
         # o[=1] = 20 #outliner
         y = m*x + c + n + o
         n = len(x)
         X = np.concatenate([x.reshape(n, 1), np.ones((n, 1))], axis=1)
         B = np.linalg.pinv(X.T@X) @ X.T @ y
         mstar = B[0]
         cstar = B[1]
         plt.plot(x, y, '+', label="Noisy Points")
         plt.plot([x[0], x[-1]], [m*x[0]+c, m*x[-1]+c],
                  color='g', linewidth=2, label=r"True Line")
         plt.plot([x[0], x[-1]], [mstar*x[0]+cstar, mstar*x[-1]+cstar],
                  color='r', linewidth=2, label=r"Estimated Line")
         plt.legend()
```

Out[]: <matplotlib.legend.Legend at 0x25cd076adf0>



```
In [ ]:
         # Least square line fitting
         m = 2
         c = 1
         x = np.arange(1, 100, 1)
         np.random.seed(45)
         sigma = 20
         n = sigma*np.random.randn(len(x))
         o = np.zeros(x.shape)
         # o[=1] = 20 #outliner
         y = m*x + c + n + o
         n = len(x)
         u11 = np.sum((x-np.mean(x))**2)
         u12 = np.sum((x-np.mean(x))*(y-np.mean(y)))
         u21 = u12
         u22 = np.sum((y-np.mean(y))**2)
         U = np.array([[u11, u12], [u21, u22]])
         W, V = np.linalg.eig(U)
         ev_corresponding_to_smallest_ev = V[:, np.argmin(W)]
         [a,b] = ev_corresponding_to_smallest_ev
         d = a*np.mean(x) + b*np.mean(y)
         mstar = -a/b
         cstar = d/b
         plt.plot(x, y, '+', label="Noisy Points")
         plt.plot([x[0], x[-1]], [m*x[0]+c, m*x[-1]+c],
                  color='g', linewidth=2, label="True Line")
         plt.plot([x[0], x[-1]], [mstar*x[0]+cstar, mstar*x[-1]+cstar],
                  color='r', linewidth=2, label="Estimated Line")
         plt.grid(True)
         plt.legend()
```

Out[]: <matplotlib.legend.Legend at 0x1be582e8f70>

