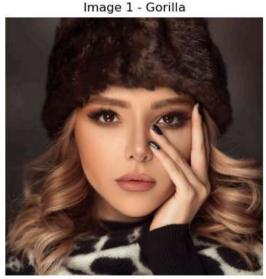
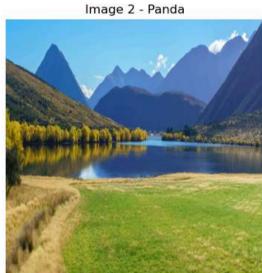
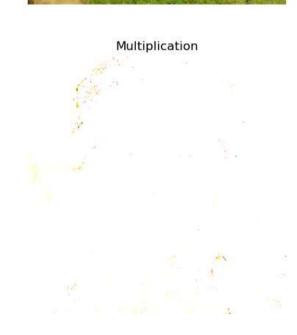
```
In [4]:
 1 import cv2
  2 import numpy as np
  3 import matplotlib.pyplot as plt
  5 img1 = cv2.imread("girl.jpg")
 7 img2 = cv2.imread("scene.jpg")
 8
 9 img2 = cv2.resize(img2, (img1.shape[1], img1.shape[0]))
 10
 11 | add = cv2.add(img1, img2)
 12 | sub = cv2.subtract(img1, img2)
 13 | mul = cv2.multiply(img1, img2)
 14 div = cv2.divide(img1, img2)
 15
 16 rows, cols = img1.shape[:2]
 17 M_trans = np.float32([[1, 0, 100], [0, 1, 50]])
 translation = cv2.warpAffine(img1, M_trans, (cols, rows))
 20 M_rot = cv2.getRotationMatrix2D((cols / 2, rows / 2), 90, 1)
 21 rotation = cv2.warpAffine(img1, M_rot, (cols, rows))
 22
 23 transform_images = [translation, rotation]
 24 | transform_titles = ['Translation', 'Rotation 90°']
 25
 26 def convert_bgr_to_rgb(img):
 27
        return cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
 28
 29 | images = [img1, img2, add, sub, mul, div]
    titles = ['Image 1 - Gorilla', 'Image 2 - Panda', 'Addition', 'Subtraction', 'Multiplication', 'Division']
 30
 31
 32 # Plot the images
 33 plt.figure(figsize=(18, 10))
 34 | for i, (img, title) in enumerate(zip(images, titles)):
 35
        plt.subplot(2, 3, i+1)
 36
        plt.imshow(convert_bgr_to_rgb(img))
 37
        plt.title(title)
         plt.axis('off')
 38
 39
```

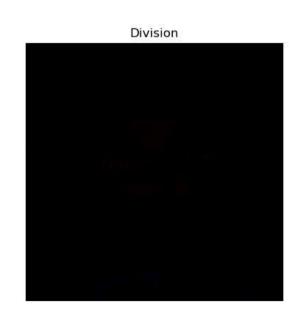












```
In [7]:
 1 import cv2
  2 import matplotlib.pyplot as plt
 4 img = cv2.imread("scene.jpg")
 6 mean = cv2.blur(img, (5, 5))
    gaussian = cv2.GaussianBlur(img, (5, 5), 0)
 8 median = cv2.medianBlur(img, 5)
 9 bilateral = cv2.bilateralFilter(img, 9, 75, 75)
10
images = [img, mean, gaussian, median, bilateral]
titles = ['original', 'mean', 'gaussian', 'median', 'bilateral']
13
plt.figure(figsize=(12, 6))
15 for i in range(5):
        plt.subplot(1, 5, i + 1)
16
        plt.imshow(cv2.cvtColor(images[i], cv2.COLOR_BGR2RGB))
17
18
        plt.title(titles[i])
19
        plt.axis('off')
20
21 plt.tight_layout()
22 plt.show()
23
```

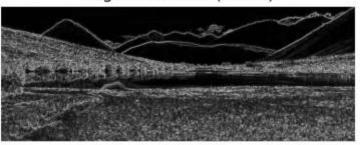


```
In [8]:
 1 import cv2
  2 import numpy as np
 3 import matplotlib.pyplot as plt
 4
 5 | # Load the image in grayscale
 6 image = cv2.imread("scene.jpg", cv2.IMREAD_GRAYSCALE)
 8 # Apply Sobel operator in x and y directions
 9 sobel_x = cv2.Sobel(image, cv2.CV_64F, 1, 0, ksize=3)
 sobel_y = cv2.Sobel(image, cv2.CV_64F, 0, 1, ksize=3)
 12 # Calculate gradient magnitude
 sobel_magnitude = np.uint8(np.absolute(cv2.magnitude(sobel_x, sobel_y)))
 14
 15 # Sharpen image by adding edge magnitude to original
 sharp_image = cv2.add(image, sobel_magnitude)
 17
 18 # Plotting the results
 19 plt.figure(figsize=(10, 10))
 20
 21 plt.subplot(1, 2, 1)
 22 plt.imshow(image, cmap='gray')
 23 plt.title("Original Image")
 24 plt.axis('off')
 25
 26 plt.subplot(1, 2, 2)
    plt.imshow(sobel_magnitude, cmap='gray')
 28 plt.title("Edge Detection (Sobel)")
 29 plt.axis('off')
 30
 31 plt.figure(figsize=(6, 6))
 32 plt.imshow(sharp_image, cmap='gray')
 33 plt.title("Sharpened Image")
 34 plt.axis('off')
 35
 36 plt.show()
 37
```

## Original Image



Edge Detection (Sobel)

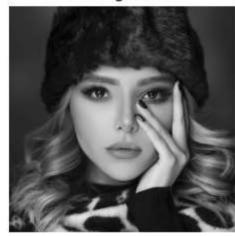


Sharpened Image



```
In [9]:
 1 import cv2
  2 import numpy as np
  3 import matplotlib.pyplot as plt
 5 # Load the image in grayscale
  6 image = cv2.imread("girl.jpg", cv2.IMREAD_GRAYSCALE)
 8 # Apply Gaussian blur before Canny to reduce noise
 9 edges = cv2.Canny(cv2.GaussianBlur(image, (5, 5), 1.5), 50, 150)
 10
 11 | # Plot original and edge-detected images
 12 plt.subplot(1, 2, 1)
 plt.imshow(image, cmap='gray')
 14 | plt.title('Original')
15 plt.axis('off')
17 plt.subplot(1, 2, 2)
18 plt.imshow(edges, cmap='gray')
 19 plt.title('Canny Edges')
 20 plt.axis('off')
 21
 22 plt.show()
 23
```

## Original



## Canny Edges



```
In [13]:
  1 import cv2
   2 | import numpy as np
   3 import matplotlib.pyplot as plt
   4 image = cv2.imread("girl.jpg", cv2.IMREAD_GRAYSCALE)
   5 | equalized = cv2.equalizeHist(image)
   6 | stretched = np.uint8(((image - np.min(image)) * 255) / (np.max(image) - np.min(image)))
     for i, (img, title) in enumerate([...]):
   8
         plt.subplot(2, 3, 2*i + 1)
  9
         plt.imshow(img, cmap='gray')
  10
  11
         plt.subplot(2, 3, 2*i + 2)
  12
         plt.hist(img.ravel(), bins=256, range=(0, 256))
  13
```

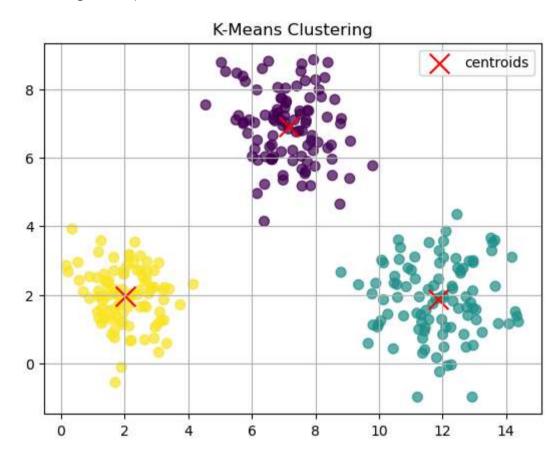
TypeError: cannot unpack non-iterable ellipsis object

```
In [20]:
   1 import cv2
     import numpy as np
   3 import matplotlib.pyplot as plt
   4 from sklearn.cluster import KMeans
   5 X = np.vstack([
         np.random.normal(loc=[2, 2], scale=0.8, size=(100, 2)),
         np.random.normal(loc=[7, 7], scale=1.0, size=(100, 2)),
         np.random.normal(loc=[12, 2], scale=1.2, size=(100, 2))
   8
     ])
  9
  10
     kmeans = KMeans(n_clusters=3, random_state=0).fit(X)
     plt.scatter(X[:, 0], X[:, 1], c=kmeans.labels_, cmap='viridis', s=50,alpha=0.7)
     plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1],c='red',marker='x',s=200,label ="centroids")
  13
  14 | plt.title('K-Means Clustering')
  15 plt.legend()
  16 plt.grid(True)
  17 plt.show()
```

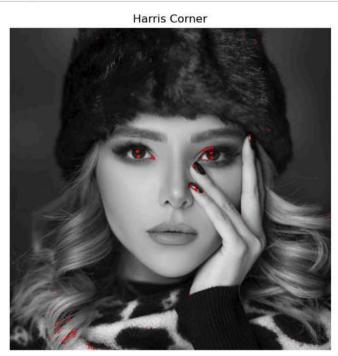
E:\Anaconda3\envs\tensorflow\Lib\site-packages\sklearn\cluster\\_kmeans.py:1412: FutureWarning: The default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the value of `n\_init` explicitly to suppress the warning super().\_check\_params\_vs\_input(X, default\_n\_init=10)

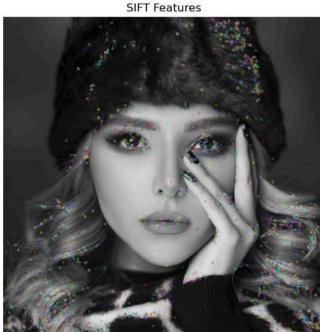
E:\Anaconda3\envs\tensorflow\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OM P\_NUM\_THREADS=2.

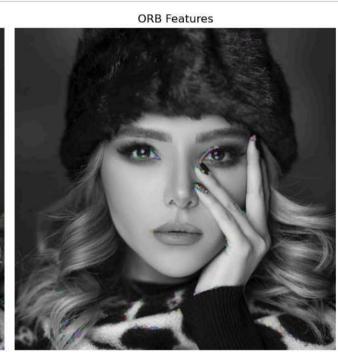
warnings.warn(



```
In [21]:
  1 | import cv2
   2 | import numpy as np
  3 | import matplotlib.pyplot as plt
  5 | # Load image in grayscale
   6 | image = cv2.imread("girl.jpg", cv2.IMREAD_GRAYSCALE)
  8 # Harris Corner Detection
  9 harris = cv2.cornerHarris(image, 2, 3, 0.04)
  10 harris = cv2.dilate(harris, None)
  image_harris = cv2.cvtColor(image, cv2.COLOR_GRAY2BGR)
  12 | image_harris[harris > 0.01 * harris.max()] = [0, 0, 255]
  13
  14 | # SIFT Feature Detection
  15 | sift = cv2.SIFT_create()
  16 kp_sift, des_sift = sift.detectAndCompute(image, None)
  image_sift = cv2.drawKeypoints(image, kp_sift, None)
  18
  19 # ORB Feature Detection
  20 orb = cv2.ORB_create()
  21 kp_orb, des_orb = orb.detectAndCompute(image, None)
  22 | image_orb = cv2.drawKeypoints(image, kp_orb, None)
  23
  24 | # Combine images and titles
  25 | images = [image_harris, image_sift, image_orb]
  titles = ['Harris Corner', 'SIFT Features', 'ORB Features']
  27
  28 # Plot results
  29 plt.figure(figsize=(15, 8))
  30 | for i, (img, title) in enumerate(zip(images, titles), 1):
         plt.subplot(1, 3, i)
  31
  32
         plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
  33
         plt.title(title)
  34
         plt.axis('off')
  35
  36 plt.tight_layout()
  37
     plt.show()
  38
  39 # Print number of keypoints
 40 print(f"SIFT Keypoints: {len(kp_sift)} | ORB Keypoints: {len(kp_orb)}")
  41
```



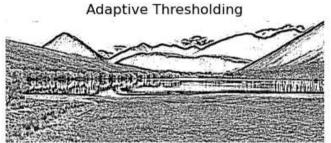




SIFT Keypoints: 2016 | ORB Keypoints: 500

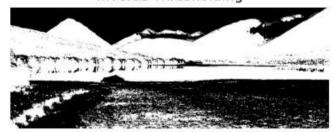
```
In [22]:
  1 import cv2
  2 import matplotlib.pyplot as plt
  4 # Load image in grayscale
  5 image = cv2.imread("scene.jpg", cv2.IMREAD_GRAYSCALE)
  7 # Apply various thresholding techniques
  8 global_thresh = cv2.threshold(image, 127, 255, cv2.THRESH_BINARY)[1]
  9 adaptive_thresh = cv2.adaptiveThreshold(image, 255, cv2.ADAPTIVE_THRESH_MEAN_C,
 10
                                            cv2.THRESH_BINARY, 11, 2)
 otsu_thresh = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)[1]
 inverse_thresh = cv2.threshold(image, 127, 255, cv2.THRESH_BINARY_INV)[1]
 13
 14 # Titles and images for display
 titles = ['Global Thresholding', 'Adaptive Thresholding', "Otsu's Thresholding", 'Inverse Thresholding']
 images = [global_thresh, adaptive_thresh, otsu_thresh, inverse_thresh]
 17
 18 # Display the images
 19 plt.figure(figsize=(12, 8))
 20 | for i, (title, img) in enumerate(zip(titles, images)):
 21
         plt.subplot(2, 3, i + 1)
         plt.imshow(img, cmap='gray')
 22
 23
         plt.title(title)
 24
         plt.axis('off')
 25
 26 plt.tight_layout()
 27 plt.show()
 28
```





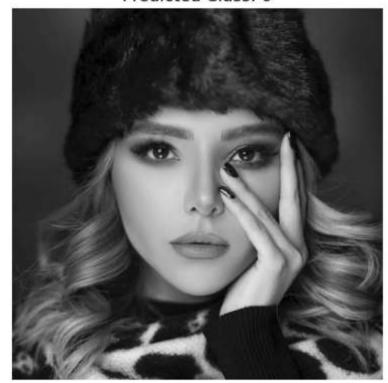


Inverse Thresholding



```
In [23]:
  1 import cv2
  2 import numpy as np
  3 from sklearn.neighbors import KNeighborsClassifier
  4 | import matplotlib.pyplot as plt
  6 # Load grayscale image
  7 image = cv2.imread("girl.jpg", cv2.IMREAD_GRAYSCALE)
  9 # Create a second image filled with zeros, same shape as the original
  10 image2 = np.zeros_like(image)
  12 # Flatten both images and use them as features
  13 X = np.array([image.flatten(), image2.flatten()])
 14 y = [0, 1] # Labels for classification
  15
  16 | # Initialize and train the K-Nearest Neighbors classifier
  17 knn = KNeighborsClassifier(n_neighbors=1)
  18 knn.fit(X, y)
  19
  20 # Predict the class of the input image
  21 | y_pred = knn.predict([image.flatten()])
 22
 23 # Display the image with predicted class
  24 plt.imshow(image, cmap='gray')
  25 plt.title(f'Predicted Class: {y_pred[0]}')
  26 plt.axis('off')
  27 plt.show()
  28
```

## Predicted Class: 0



```
In [25]:
  1 import cv2
  2 import numpy as np
  3 import matplotlib.pyplot as plt
  5 img_local = cv2.imread("scene.jpg")
  7 gray_local = cv2.cvtColor(img_local, cv2.COLOR_BGR2GRAY)
  8 kernel = np.ones((5, 5), np.float32) / 25
  9 local_corr = cv2.filter2D(gray_local, -1, kernel)
 10
 11 | images = [gray_local, local_corr]
 titles = ['Original Grayscale Image', 'Local Correlation Image']
 cmaps = ['gray', 'inferno']
 14
 plt.figure(figsize=(12, 6))
 16 | for i, (img, title, cmap) in enumerate(zip(images, titles, cmaps), 1):
 17
         plt.subplot(1, 2, i)
 18
         plt.imshow(img, cmap=cmap)
 19
         plt.title(title)
 20
         plt.axis('off')
 21
 22 plt.suptitle('Original vs Local Correlation', fontsize=16)
 23 plt.tight_layout()
 24 plt.show()
 25
```

Original vs Local Correlation



