This code is for scanning an image of a document and processing it to obtain a clean version of the document with minimal noise. Let's go through it step by step:

Step -1 Getting Bird's Eye View of the Image

```
def scan(img):
   # Resize image to workable size
    dim limit = 1080
   max dim = max(img.shape)
    if max dim > dim limit:
        resize scale = dim limit / max dim
        img = cv2.resize(img, None, fx=resize scale, fy=resize scale)
    # Create a copy of resized original image for later use
    orig img = img.copy()
    # Repeated Closing operation to remove text from the document.
    kernel = np.ones((5, 5), np.uint8)
    img = cv2.morphologyEx(img, cv2.MORPH CLOSE, kernel, iterations=3)
    # GrabCut
    mask = np.zeros(img.shape[:2], np.uint8)
    bgdModel = np.zeros((1, 65), np.float64)
    fgdModel = np.zeros((1, 65), np.float64)
    rect = (20, 20, img.shape[1] - 20, img.shape[0] - 20)
    cv2.grabCut(img, mask, rect, bgdModel, fgdModel, 5, cv2.GC INIT WITH RECT)
    mask2 = np.where((mask == 2) | (mask == 0), 0, 1).astype('uint8')
    img = img * mask2[:, :, np.newaxis]
    gray = cv2.cvtColor(img, cv2.COLOR BGR2GRAY)
    gray = cv2.GaussianBlur(gray, (11, 11), 0)
    # Edge Detection.
    canny = cv2.Canny(gray, 0, 200)
   canny = cv2.dilate(canny, cv2.getStructuringElement(cv2.MORPH ELLIPSE, (5,
5)))
    # Finding contours for the detected edges.
    contour, hierarchy = cv2.findContours(canny, cv2.RETR LIST,
cv2.CHAIN APPROX NONE)
    # Keeping only the largest detected contour.
    page = sorted(contour, key=cv2.contourArea, reverse=True)[:5]
   # Detecting Edges through Contour approximation.
   # Loop over the contours.
    if len(page) == 0:
        return orig img
    for c in page:
        # Approximate the contour.
        epsilon = 0.02 * cv2.arcLength(c, True)
        corners = cv2.approxPolyDP(c, epsilon, True)
       # If our approximated contour has four points.
        if len(corners) == 4:
            break
```

```
dim_limit = 1080
max_dim = max(img.shape)
if max_dim > dim_limit:
    resize_scale = dim_limit / max_dim
    img = cv2.resize(img, None, fx=resize_scale, fy=resize_scale)
```

The function first resizes the input image if its size is larger than a certain threshold (dim_limit). The maximum dimension of the input image is calculated and if it is greater than the threshold, the image is resized to fit within the limit while maintaining its aspect ratio. This is done to ensure that the processing is faster and more efficient.

```
orig_img = img.copy()
```

This creates a copy of the resized image for later use.

```
kernel = np.ones((5, 5), np.uint8)
img = cv2.morphologyEx(img, cv2.MORPH_CLOSE, kernel, iterations=3)
```

The function then applies a morphological operation called "Closing" to the input image. Closing is a dilation followed by an erosion operation and is commonly used to remove noise and fill gaps in an image. In this case, it is used to remove any text from the document. A kernel of size 5x5 is used for this operation and it is applied three times.

```
mask = np.zeros(img.shape[:2], np.uint8)
bgdModel = np.zeros((1, 65), np.float64)
fgdModel = np.zeros((1, 65), np.float64)
rect = (20, 20, img.shape[1] - 20, img.shape[0] - 20)
cv2.grabCut(img, mask, rect, bgdModel, fgdModel, 5, cv2.GC_INIT_WITH_RECT)
mask2 = np.where((mask == 2) | (mask == 0), 0, 1).astype('uint8')
img = img * mask2[:, :, np.newaxis]
```

The function then uses the GrabCut algorithm to separate the foreground (document) from the background. GrabCut is an iterative segmentation algorithm that uses a user-defined bounding box to separate the foreground from the background. The algorithm updates the segmentation iteratively based on the user input. In this case, a rectangular bounding box is defined using the dimensions of the input image with some margin. The GrabCut algorithm then segments the input image and returns a mask. The mask is then thresholded to create a binary mask, which is used to remove the background from the input image.

```
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
gray = cv2.GaussianBlur(gray, (11, 11), 0)
```

The function converts the input image to grayscale and applies a Gaussian blur to it. Gaussian blur is a common filter used to smooth an image and reduce noise. In this case, it is used to prepare the image for edge detection.

```
canny = cv2.Canny(gray, 0, 200)
canny = cv2.dilate(canny, cv2.getStructuringElement(cv2.MORPH_ELLIPSE, (5, 5)))
```

The function then applies the Canny edge detection algorithm to the blurred image to detect edges. Canny is an edge detection algorithm that uses a multi-stage process to detect a wide range of edges in an image. In this case, it is used to detect the edges

```
destination_corners = find_dest(corners)
```

```
def order points(pts):
    '''Rearrange coordinates to order:
      top-left, top-right, bottom-right, bottom-left'''
    rect = np.zeros((4, 2), dtype='float32')
    pts = np.array(pts)
    s = pts.sum(axis=1)
    # Top-left point will have the smallest sum.
    rect[0] = pts[np.argmin(s)]
    # Bottom-right point will have the largest sum.
    rect[2] = pts[np.argmax(s)]
    diff = np.diff(pts, axis=1)
    # Top-right point will have the smallest difference.
    rect[1] = pts[np.argmin(diff)]
    # Bottom-left will have the largest difference.
    rect[3] = pts[np.argmax(diff)]
    # return the ordered coordinates
    return rect.astype('int').tolist()
```

The order_points() function takes in the detected corners of the page as an argument and reorders them to have the top-left, top-right, bottom-right, and bottom-left points in a consistent order. This is important for the later steps of the image processing pipeline to work correctly.

The function first creates a 4x2 numpy array of float 32 data type called rect to store the reordered points. It then converts the input pts argument to a numpy array and computes the sum of each row of coordinates using the sum() method and the axis=1 argument. The argmin() and argmax() methods are then used to find the indices of the points that have the smallest and largest sums, respectively. The top-left point is assigned to rect[0] and the bottom-right point is assigned to rect[2].

The function then computes the difference between the x and y coordinates of the input pts array using the diff() method and the axis=1 argument. The argmin() and argmax() methods are again used to find the indices of the points that have the smallest and largest differences, respectively. The top-right point is assigned to rect[1] and the bottom-left point is assigned to rect[3].

Finally, the function returns the reordered rect array as a list of integers using the astype() method and the 'int' argument.

```
def find_dest(pts):
    (tl, tr, br, bl) = pts
    # Finding the maximum width.
    widthA = np.sqrt(((br[0] - bl[0]) ** 2) + ((br[1] - bl[1]) ** 2))
    widthB = np.sqrt(((tr[0] - tl[0]) ** 2) + ((tr[1] - tl[1]) ** 2))
    maxWidth = max(int(widthA), int(widthB))

# Finding the maximum height.
heightA = np.sqrt(((tr[0] - br[0]) ** 2) + ((tr[1] - br[1]) ** 2))
heightB = np.sqrt(((tl[0] - bl[0]) ** 2) + ((tl[1] - bl[1]) ** 2))
maxHeight = max(int(heightA), int(heightB))
# Final destination co-ordinates.
destination_corners = [[0, 0], [maxWidth, 0], [maxWidth, maxHeight], [0, maxHeight]]

return order_points(destination_corners)

destination_corners = find_dest(corners)
```

The find_dest(pts) function calculates the final destination points of the transformed image.

The function takes in the four corner points of the detected document as <code>pts</code>. These points are assumed to be in an unordered format. The function then rearranges these points using the <code>order_points(pts)</code> function defined earlier to get the points in the order of the top-left, top-right, bottom-right, and bottom-left.

Once the points are ordered, the function calculates the maximum width and height of the document using the Euclidean distance formula between the corner points.

widthA calculates the width of the document using the distance between the bottom-right corner br and the bottom-left corner bl. Similarly, widthB calculates the width of the document using the distance between the top-right corner tr and the top-left corner tl. The maxWidth variable is then set to the maximum of these two values.

heightA calculates the height of the document using the distance between the top-right corner tr and the bottom-right corner br. Similarly, heightB calculates the height of the document using the distance between the top-left corner tl and the bottom-left corner bl. The maxHeight variable is then set to the maximum of these two values.

Finally, the destination_corners list is created using the maxWidth and maxHeight values calculated earlier. These points are in the order of top-left, top-right, bottom-right, and bottom-

left. The order_points(destination_corners) function is then called to order the points correctly and the ordered coordinates of the final destination points are returned.

The code above performs the final transformation of the original image using the perspective transformation matrix obtained from the homography.

First, the height and width of the original image are extracted using the shape attribute of the NumPy array orig_img. Then, the getPerspectiveTransform() function of OpenCV is used to get the homography matrix M. This function takes in the four corner points of the source image (corners) and their corresponding four corner points in the destination image (destination_corners), and returns the homography matrix M.

Finally, the <code>warpPerspective()</code> function of OpenCV is used to perform the actual perspective transformation. It takes in the original image <code>orig_img</code>, the homography matrix <code>M</code>, and the output size of the destination image (the width and height of the bottom-right corner point of the destination image) as input. The <code>flags</code> parameter is set to <code>cv2.INTER_LINEAR</code>, which specifies the interpolation method used to perform the transformation.

The output of this function is the final transformed image final, which has been cropped and warped to appear as though it was taken from a bird's-eye view perspective.

Step-2 Tuning the Image

```
def increase brightness(img, value=30):
    hsv = cv2.cvtColor(img, cv2.COLOR BGR2HSV)
    h, s, v = cv2.split(hsv)
    lim = 255 - value
    v[v > lim] = 255
    v[v <= lim] += value
    final hsv = cv2.merge((h, s, v))
    img = cv2.cvtColor(final hsv, cv2.COLOR HSV2BGR)
    return img
def final image(rotated):
# Create our shapening kernel, it must equal to one eventually
  kernel sharpening = np.array([0,-1,0],
                                [-1, 5, -1],
                                [0, -1, 0]]
  # applying the sharpening kernel to the input image & displaying it.
  sharpened = cv2.filter2D(rotated, -1, kernel sharpening)
  sharpened=increase brightness(sharpened, 35)
  return sharpened
```

increase_brightness(img, value=30):

- This function takes an image img and an optional value parameter to increase the brightness.
- hsv = cv2.cvtColor(img, cv2.CoLOR_BGR2HSV): The image is first converted from the BGR color space to the HSV color space.
- h, s, v = cv2.split(hsv): The split() function is used to separate the hue, saturation, and value channels of the image.
- lim = 255 value: lim is a threshold value for the brightness.
- v[v > lim] = 255: If the pixel value in the value channel is greater than lim, set it to the maximum value of 255.
- v[v <= lim] += value: If the pixel value in the value channel is less than or equal to lim, add the value parameter to it.
- final_hsv = cv2.merge((h, s, v)): The merge() function is used to merge the
 modified hue, saturation, and value channels back into a single image in the HSV color
 space.
- img = cv2.cvtColor(final_hsv, cv2.CoLOR_HSV2BGR): Finally, the image is converted back to the BGR color space.

final_image(rotated):

- This function takes a rotated image as input.
- kernel_sharpening = np.array([[0,-1,0], [-1, 5,-1], [0,-1,0]]): A sharpening kernel is created as a 3x3 numpy array with the center element having the highest weight (5).
- sharpened = cv2.filter2D(rotated, -1, kernel_sharpening): The sharpening kernel is applied to the input image using the filter2D() function in OpenCV.
- sharpened=increase_brightness(sharpened, 35): The increase_brightness() function is called to increase the brightness of the image.
- return sharpened: The final sharpened and brightened image is returned.

```
cleaned_image = final_image(scan(img))
cv2.immshow("Image",cleaned_image)
cv2.waitKey()
```

Finally show the Image.