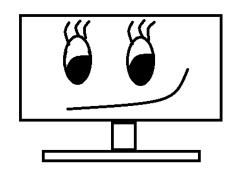
Seneca



CVI620/ DP\$920 Introduction to Computer Vision

Image Morphology & Geometric Transformations

Seneca College

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Overview

- Morphology
- Binary Masking
- Geometric Transformation

Morphology

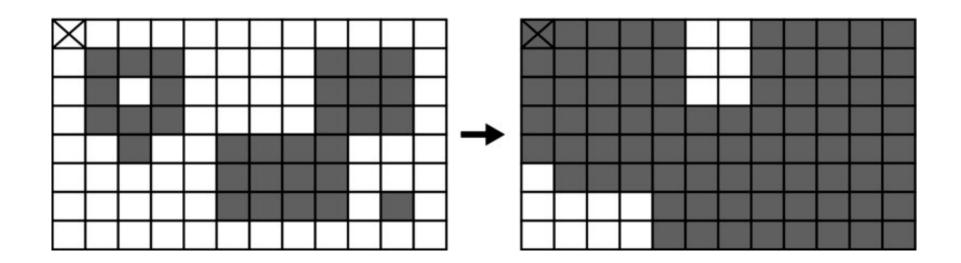
The study of structure or form [Merriam-Webster]

Morphology [2]

- Nonlinear filtering, often applied to binary images
- Kernels: often a square kernel or a disc, but can be more complicated
- Basic operations:
 - Dilate: Convolution of some image with a kernel in which any given pixel is replaced with the *local maximum* of all pixel values covered by the kernel
 - Effect: causes filled regions to grow
 - Good for finding connected components
 - Erode: Convolution of some image with a kernel in which any given pixel is replaced with the *local minimum* of all pixel values covered by the kernel
 - Good for removing smaller areas (noise)

Example: Dilate a binary image [3]

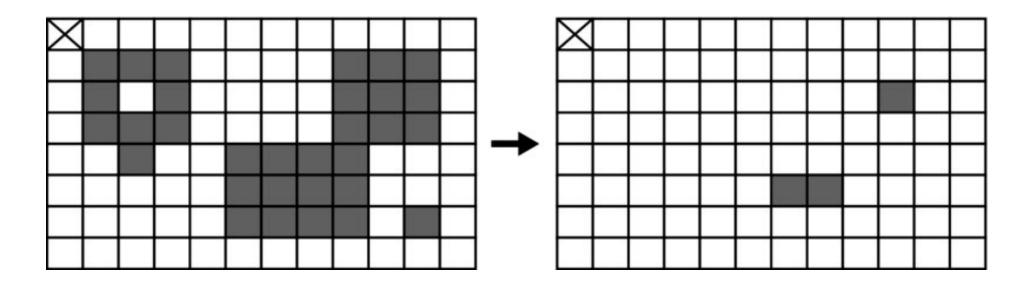
Assuming white is zero in the following image



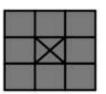
3x3 square kernel:

Example: Erode a binary image [3]

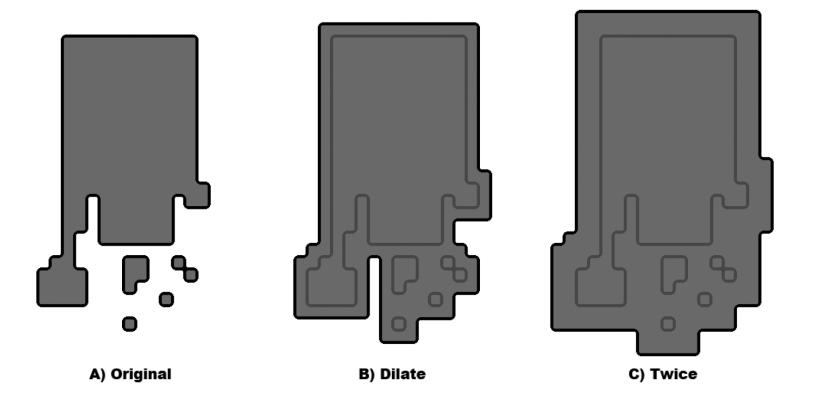
Assuming white is zero in the following image



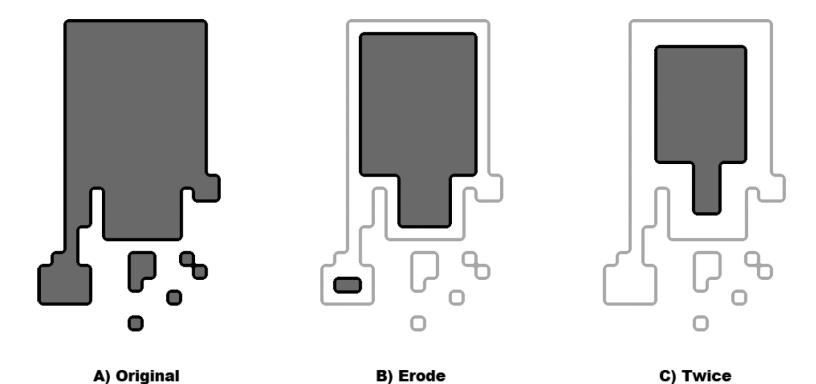
3x3 square kernel:



```
void cv::dilate(
 cv::InputArray
                                   // Input image
                  SCC,
 cv::OutputArray
                  dst,
                                   // Result image
 cv::InputArray
                 element,
                                                 // Structuring, a cv::Mat()
           anchor = cv::Point(-1,-1), // Location of anchor point
 cv::Point
                  iterations = 1, // Number of times to apply
 int
                  borderType = cv::BORDER_CONSTANT // Border extrapolation
 int
 const cv::Scalar& borderValue = cv::morphologyDefaultBorderValue()
);
```



```
void cv::erode(
 cv::InputArray
                                                      Input image
                   SCC,
                                                    // Result image
 cv::OutputArray
                   dst,
 cv::InputArray
                   element,
                                                   // Structuring, a cv::Mat()
                                                  // Location of anchor point
                              = cv::Point(-1,-1),
 cv::Point
                   anchor
                   iterations = 1,
                                   // Number of times to apply
 int
                   borderType = cv::BORDER_CONSTANT // Border extrapolation
 int
 const cv::Scalar& borderValue = cv::morphologyDefaultBorderValue()
);
```

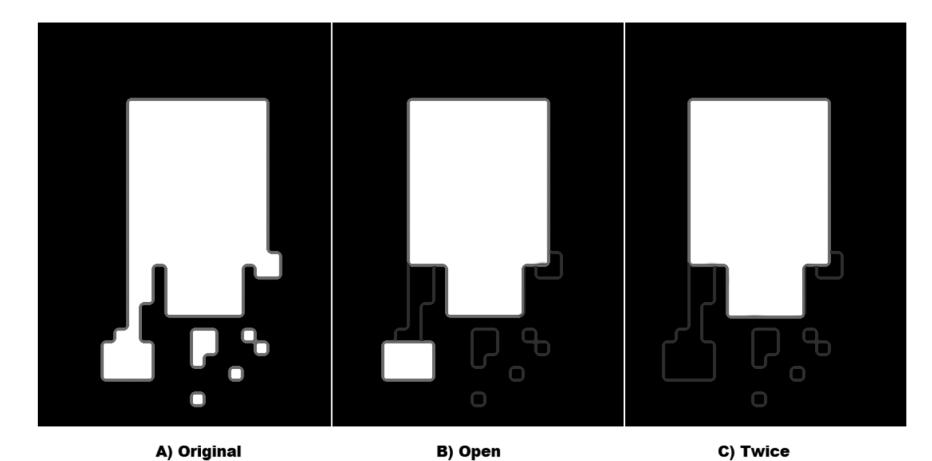


Opening

Example:

cv::Mat elem = 255 * cv::Mat::ones(s, s, CV_8UC1);
cv::morphologyEx(binI, morph, MORPH_OPEN, elem);

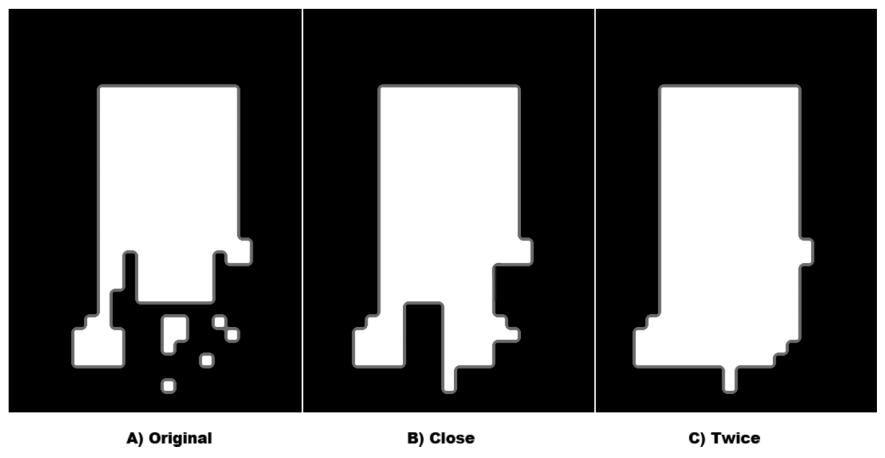
- Erode first, then dilate
- For example, for counting objects in a binary image



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Closing

• Dilate first, then erode



```
void cv::morphologyEx(
                                            // Input image
 cv::InputArray
                  SCC,
 cv::OutputArray
                  dst,
                                            // Result image
 int
                                            // Operator (e.g. cv::MOP_OPEN)
                  op,
                                            // Structuring element, cv::Mat()
                  element,
 cv::InputArray
                              = cv::Point(-1,-1), // Location of anchor point
 cv::Point
            anchor
                  iterations = 1, // Number of times to apply
 int
 int
                  borderType = cv::BORDER DEFAULT // Border extrapolation
 const cv::Scalar& borderValue = cv::morphologyDefaultBorderValue()
);
```

Value of operation	Morphological operator	Requires temp image?
cv::MOP_OPEN	Opening	No
cv::MOP_CLOSE	Closing	No
cv::MOP_GRADIENT	Morphological gradient	Always
cv::MOP_TOPHAT	Top Hat	For in-place only (src = dst)
cv::MOP_BLACKHAT	Black Hat	For in-place only (src = dst)

type of morphological operation

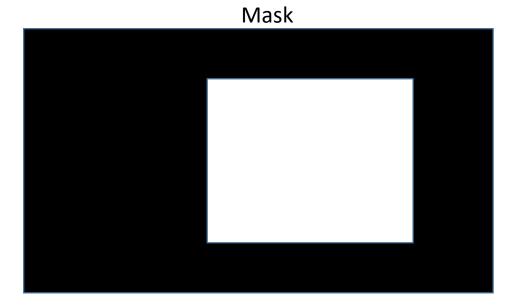
Enumerator			
MORPH_ERODE	see cv::erode		
MORPH_DILATE	see cv::dilate		
MORPH_OPEN	an opening operation		
	$\mathtt{dst} = \mathrm{open}(\mathtt{src}, \mathtt{element}) = \mathrm{dilate}(\mathrm{erode}(\mathtt{src}, \mathtt{element}))$		
MORPH_CLOSE	a closing operation		
	$\mathtt{dst} = \operatorname{close}(\mathtt{src}, \mathtt{element}) = \operatorname{erode}(\operatorname{dilate}(\mathtt{src}, \mathtt{element}))$		
MORPH_GRADIENT	a morphological gradient		
	${\tt dst} = {\tt morph_grad}({\tt src}, {\tt element}) = {\tt dilate}({\tt src}, {\tt element}) - {\tt erode}({\tt src}, {\tt element})$		
MORPH_TOPHAT	"top hat"		
	$\mathtt{dst} = \mathtt{tophat}(\mathtt{src}, \mathtt{element}) = \mathtt{src} - \mathtt{open}(\mathtt{src}, \mathtt{element})$		
MORPH_BLACKHAT	"black hat"		
	$\mathtt{dst} = \mathrm{blackhat}(\mathtt{src}, \mathtt{element}) = \mathrm{close}(\mathtt{src}, \mathtt{element}) - \mathtt{src}$		

Binary Masking

Binary Masking

- Binary Mask:
 - A matrix of 0's and 1's, with the same size as the image
 - Only compute the function/ apply the changes for nonzero elements





Masking in OpenCV

Examples in OpenCV:

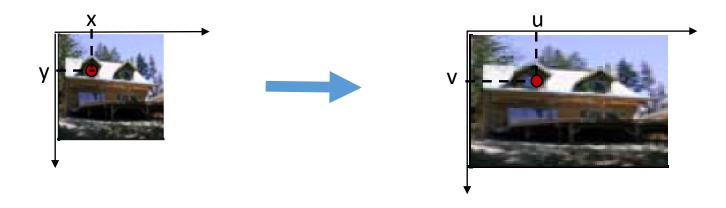
```
M0.copyTo( m1, mask ); Same as m0.copyTo(m1), except only entries indicated in the array mask are copied

m0.setTo( s, mask ); Set all entries in m0 to singleton value s; if mask is present, set only those values corresponding to nonzero elements in mask
```

Geometric Transformation

2D Transformations

 A pixel in the source image at location (x,y) is mapped to location (u,v) in the destination image



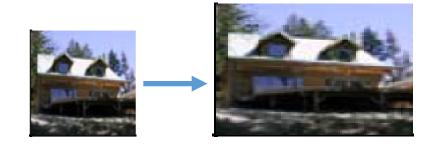
Types of 2D Transformation

- Translation pixels move in the same direction
 - $u = x + t_x$
 - $v = y + t_y$

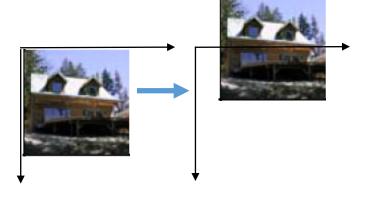


•
$$u = x * s_x$$

•
$$v = y * s_v$$



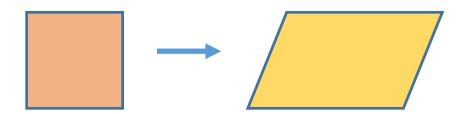
- Rotation
 - $u = x * \cos \theta y * \sin \theta$
 - $v = y * \sin \theta + x * \cos \theta$





Types of 2D Transformation (cont.)

- Shear
 - $u = x + y * sh_x$
 - $u = y + x * sh_y$
 - If $sh_y = 0$



Matrix Notation for Affine Transformations

• Affine:

• Translation:

• Scale/ Resize:

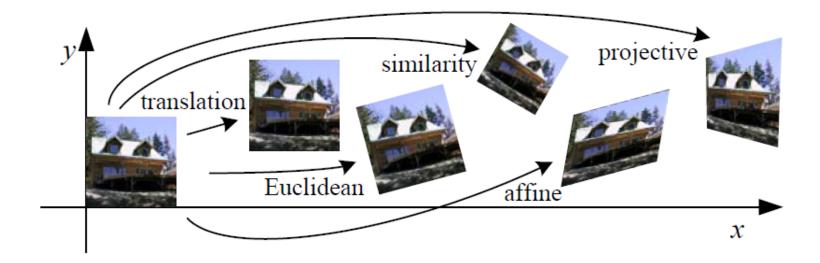
- Rotation:
- Shear:

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} a & b & t_x \\ c & d & t_y \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

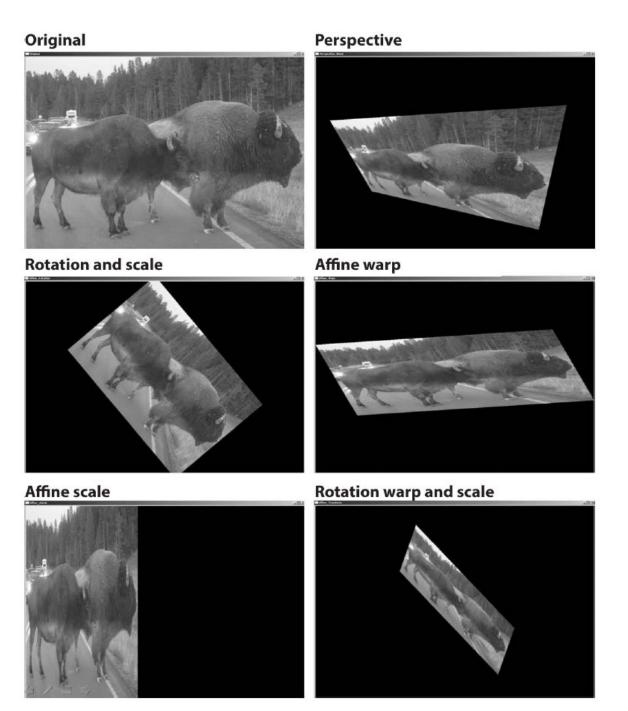
$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} s_{\chi} & 0 & 0 \\ 0 & s_{y} & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$
$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} 1 & sh_x & 0 \\ sh_y & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

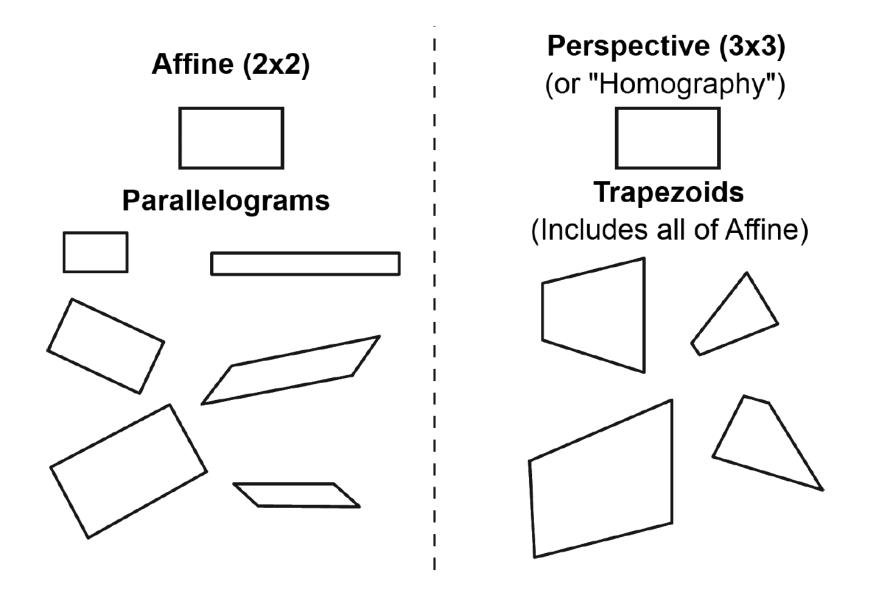


2D Geometric Image Transformations

Transformation	Matrix	# DoF	Preserves	Icon
translation	$\left[egin{array}{c c} I & t \end{array} ight]_{2 imes 3}$	2	orientation	
rigid (Euclidean)	$\left[egin{array}{c c} oldsymbol{R} & oldsymbol{t} \end{array} ight]_{2 imes 3}$	3	lengths	\Diamond
similarity	$\left[\begin{array}{c c} sR & t\end{array}\right]_{2 \times 3}$	4	angles	\Diamond
affine	$\left[egin{array}{c} oldsymbol{A} \end{array} ight]_{2 imes 3}$	6	parallelism	
projective	$\left[egin{array}{c} ilde{m{H}} \end{array} ight]_{3 imes 3}$	8	straight lines	



CVI620/DPS920- Image Morphology & Geometric Transformations



Implementations:

- Forward implementation:
 - For every pixel in the source image at location (x,y)
 - Calculate (u,v)
 - Copy the pixel: $I_{dst}(u, v) = I_{src}(x, y)$
- Backward implementation
 - For every pixel in the destination image at location (u,v)
 - Calculate (x,y) inverse transformation
 - Estimate the value of pixel at (x,y)
 - Copy the pixel: $I_{dst}(u, v) = I_{src}(x, y)$

Affine Transform using OpenCV

• Given the 2x3 transform matrix M, find the result dst

```
void cv::warpAffine(
 cv::InputArray
                                                     // Input image
                   SCC.
                                                     // Result image
 cv::OutputArray
                   dst,
 cv::InputArray
                                                     // 2-by-3 transform mtx
                   Μ,
 cv::Size
                                                     // Destination image size
                   dsize,
                   flags = cv::INTER_LINEAR, // Interpolation, inverse
 int
                   borderMode = cv::BORDER_CONSTANT, // Pixel extrapolation
 int
 const cv::Scalar& borderValue = cv::Scalar()
                                                     // For constant borders
```

Find the transform

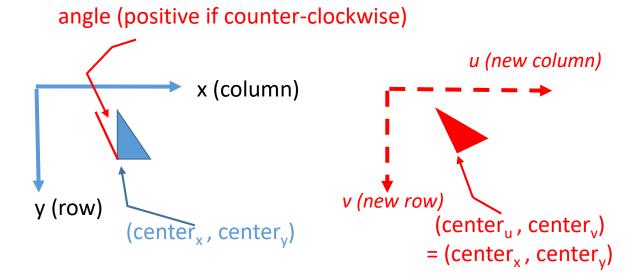
• Given the resulting image, find the transformation matrix

Get the similarity transform matrix

If we define $\alpha = \text{scale} * \cos(\text{angle})$ and $\beta = \text{scale} * \sin(\text{angle})$, then this function computes the matrix M to be:

$$M = \begin{bmatrix} \alpha & \beta & (1 - \alpha) \cdot center_x - \beta \cdot center_y \\ -\beta & \alpha & \beta \cdot center_x + (1 - \alpha) \cdot center_y \end{bmatrix}$$

$$\begin{bmatrix} u \\ v \end{bmatrix} = M \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$



Rotate an image

```
angle = 30;
M = getRotationMatrix2D(cv::Point2f(img.cols / 2.0, img.rows /2.0), angle, 1);
warpAffine(img, dst, M, img.size());
```



Resize an image

Also the resize function designed specifically for scaling the image

Table 11-1. cv::resize() interpolation options

Interpolation	Meaning
cv::INTER_NEAREST	Nearest neighbor
cv::INTER_LINEAR	Bilinear
cv::INTER_AREA	Pixel area resampling
cv::INTER_CUBIC	Bicubic interpolation
cv::INTER_LANCZOS4	Lanczos interpolation over 8×8 neighborhood.



It is important to notice the difference between cv::resize() and the similarly named cv::Mat::resize() member function of the cv::Mat class. cv::resize() creates a new image of a different size, over which the original pixels are mapped. The cv::Mat::resize() member function resizes the image whose member you are calling, and crops that image to the new size. Pixels are not interpolated (or extrapolated) in the case of cv::Mat::resize().

Resize an image

```
fx = 1;
fy = 2;
resize(img, dst, Size(), fx, fy);
```



Find the inverse transform

• Given the transform matrix, find the inverse

Perspective Transform

• Given the 3x3 transform matrix M, find the result dst

```
void cv::warpPerspective(
 cv::InputArray
                                                         Input image
                   SCC.
 cv::OutputArray
                   dst.
                                                      // Result image
 cv::InputArray
                                                      // 3-by-3 transform mtx
                   Μ,
 cv::Size
                   dsize,
                                                      // Destination image size
                   flags
                                                      // Interpolation, inverse
 int
                               = cv::INTER_LINEAR,
                   borderMode = cv::BORDER_CONSTANT, // Extrapolation method
 int
 const cv::Scalar& borderValue = cv::Scalar()
                                                      // For constant borders
);
```

Find the perspective transform

• Given the resulting image, find the transformation matrix

References

- [1] Computer Vision: Algorithms and Applications, R. Szeliski (http://szeliski.org/Book)
- [2] Learning OpenCV 3, A. Kaehler & G. Bradski
 - Available online through Safari Books, Seneca libraries
 - https://senecacollege-primo.hosted.exlibrisgroup.com/primoexplore/fulldisplay?docid=01SENC_ALMA5153244920003226&context=L&vid=01SENC&searc h scope=default_scope&tab=default_tab&lang=en_US
- [3] Practical introduction to Computer Vision with OpenCV, Kenneth Dawson-Howe
 - Available through Seneca libraries
 - https://senecacollege-primo.hosted.exlibrisgroup.com/primoexplore/fulldisplay?docid=01SENC_ALMA5142810950003226&context=L&vid=01SENC&s earch_scope=default_scope&tab=default_tab&lang=en_US