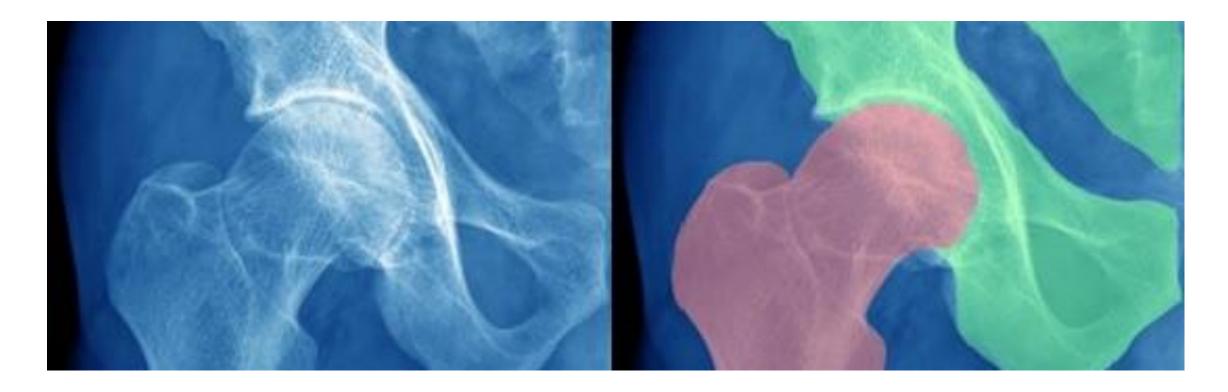


Image Segmentation

From top-down to bottom-up





Agenda

- Recap
- Image segmentation (focus on classical methods):
 - Gray-level Thresholding
 - Kernel & Filtering
 - Region Growing
 - Level set
 - Edge Detection
 - Active contours
- Segmentation Evaluation:
 - DICE Coefficient
 - JACCARD Index



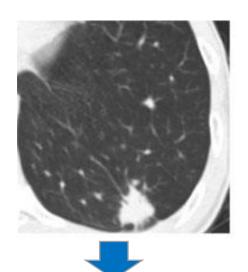
RECAP





Recall: ML Imaging Tasks

Classification

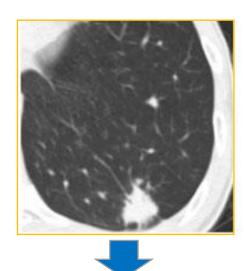


Adenocarcinoma

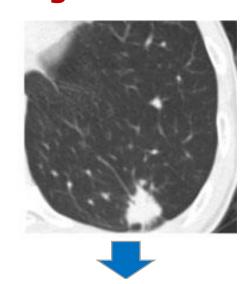
Adenoma Mesothelioma

...

Detection



Segmentation





Recall: ML Imaging Tasks

Output Space & Size?

Classification

Detection

Segmentation

Multiclass: # of Classes

Multilabel: 2^n

4 numbers: $(x_1, y_1), (x_2, y_2)$ \mathbb{R}

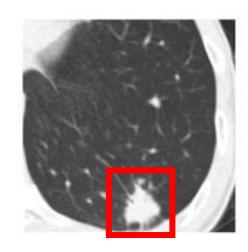
 n^{w*h}

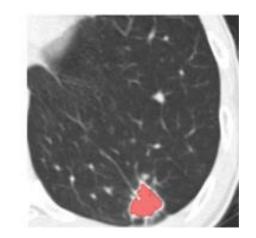


Adenoma

Mesothelioma

. . .







General Recipe for ML Modeling

TO COLUMN TO THE TOTAL PROPERTY OF THE TOTAL

Define of the problem

Collect (Historical) data

(Always!) Look at the data (e.g., analyze statistics)

Decide on an evaluation metric(s)

Define Features

Split the dataset

Train and compare Models

Hyperparameter tuning

Select a model and apply (inference / predictions)



General Recipe for ML Modeling

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Hyperparameter tuning

Select a model and apply (inference / predictions)



Decide on an evaluation metric(s)

What metrics do you know?



Decide on an evaluation metric(s)

What would you use for:							
Classification?	Classification? Regression?		Detection?	Segmentation?			



Classification?

Precision

Recall

F1-Score

Mathew's Correlation Coefficient (MCC)

ROC

AUC

Confusion Matrix



Classification?

Precision

Recall

F1-Score

Mathew's Correlation Coefficient (MCC)

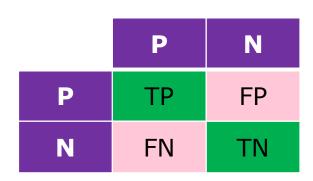
ROC

AUC

Confusion Matrix

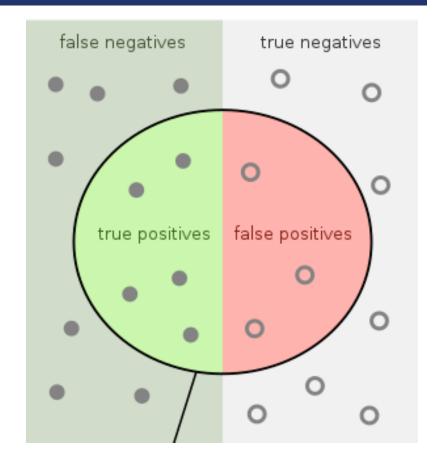


Classification



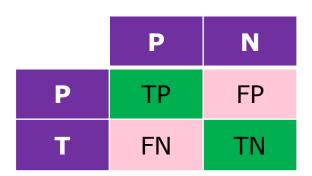
Precision:
$$\frac{TP}{TP + FP}$$

Recall:
$$\frac{TP}{TP + FN}$$



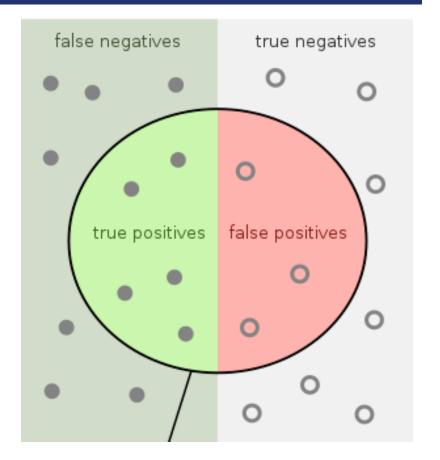


Classification



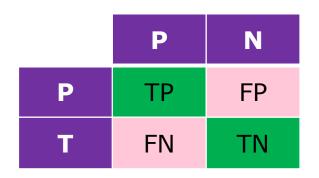
$$2\frac{recall \times precision}{recall + precision}$$





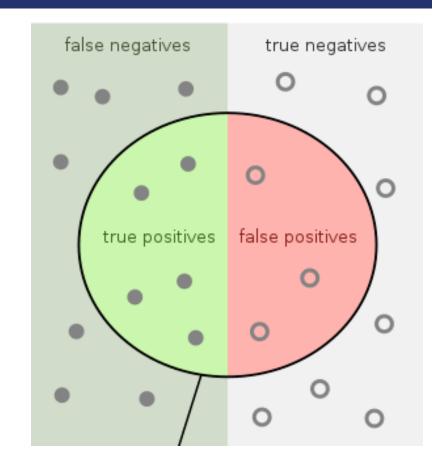


Classification



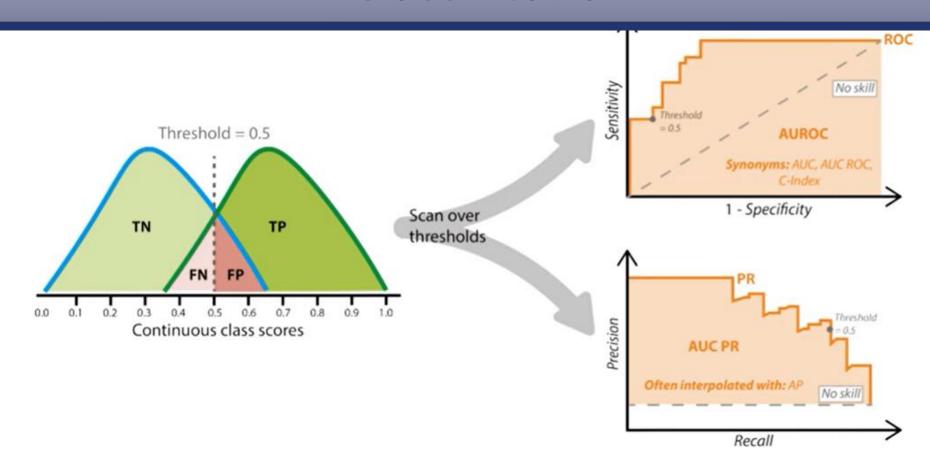
MCC:

$$\frac{TP \times TN - FP \times FN}{\sqrt{(TP + FP)(TP + FN)(TN + FP)(TN + FN)}}$$





Classification





Decide on an evaluation metric(s)

What would you use for:							
Classification?	Classification? Regression?		Detection?	Segmentation?			



Regression

$$MSE = \frac{1}{N} \sum_{i=1}^{N} (y_i - \hat{y})^2$$

$$RMSE = \sqrt{MSE}$$

$$MAE = \frac{1}{N} \sum_{i=1}^{N} |y_i - \hat{y}|$$



Decide on an evaluation metric(s)

What would you use for:							
Classification?	Classification? Regression?		Detection?	Segmentation?			



What wo

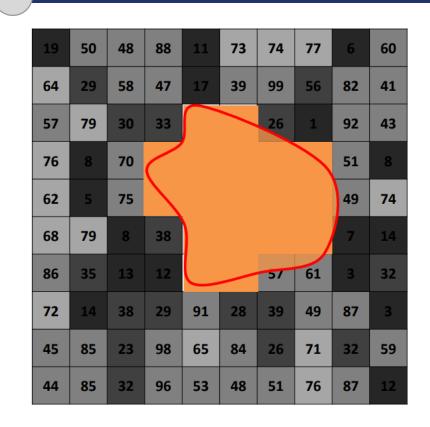
Segment

Keep in mind:

- Images are 2D, 3D, 4D...
- Most medical images are 16-bit

Definition: Clustering image elements as "belong together"

- Partitioning
 - Divide into regions/sequences with coherent internal properties
- Grouping
 - Identify sets of coherent tokens in image



65535

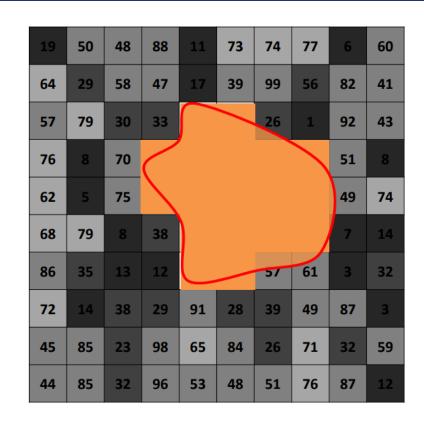
0



Segmentation

Definition: Clustering image elements as "belong together"

- Two approaches:
 - Pixel-wise categorial labels
 - Implicit Representations

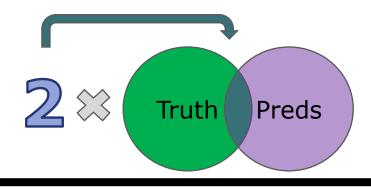


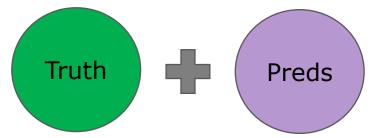




Segmentation

DICE Coefficient (F1-Score)





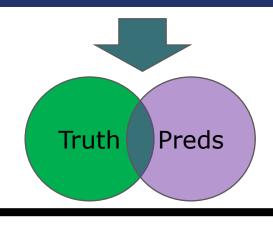
2 * Overlap Area

Total Pixel Combined

Division for normalization (to be between 0-1)



Segmentation



Jaccard Index (AUC)

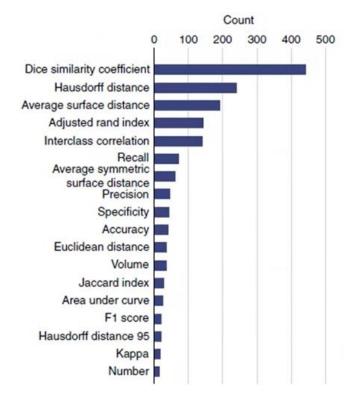
$$IOU = \frac{Overlap\ Area}{Area\ of\ Union}$$

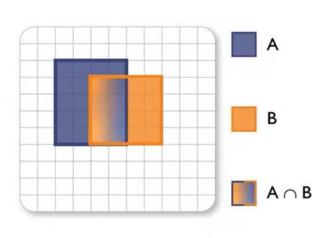
Division for normalization (to be between 0-1)



Dice Similarity Coefficient

Most widely used metric

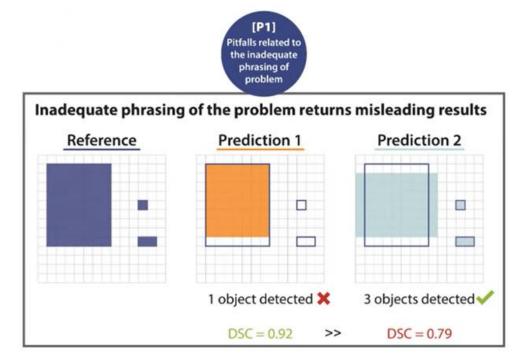






Dice Similarity Coefficient

Pitfalls: Object Detection vs. Segmentation



[2206.01653] Metrics reloaded: Recommendations for image analysis validation (arxiv.org)



Motivation

- Organ detection
 - Heart, liver, lungs in MRI / CT Scans
- Brain segmentation
 - Thalamus, hippocampus
- Quantitative Measurements
 - Ultrasound
- Image-guided surgery

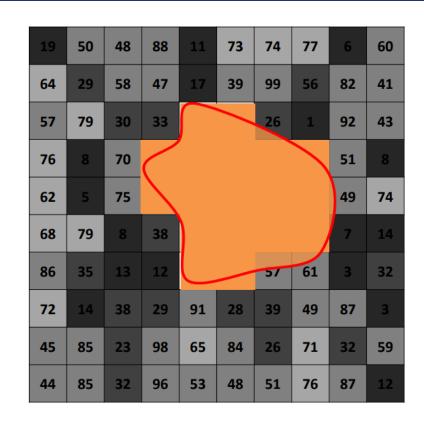




Segmentation

Definition: Clustering image elements as "belong together"

- Two approaches:
 - Pixel-wise categorial labels
 - Implicit Representations



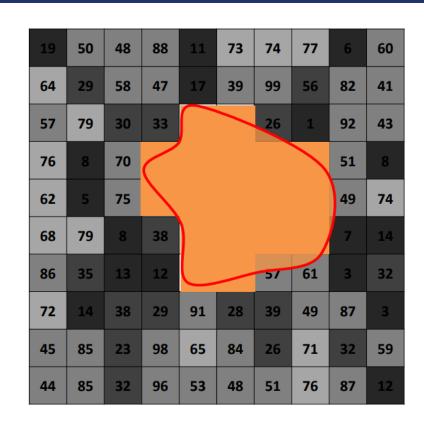




Segmentation

Definition: Clustering image elements as "belong together"

- Two approaches:
 - Pixel-wise categorial labels
 - Implicit Representations







Pixel-wise Image Segmentation

Pixel-wise labeling is the most common representation.

Categorical labels can range from 0–N.

19	50	48	88	11	73	74	77	6	60
64	29	58	47	17	39	99	56	82	41
57	79	30	33	134	145	26	1	92	43
76	8	70	100	184	173	156	176	51	8
62	5	75	118	176	189	189	163	49	74
68	79	8	38	103	127	110	164	7	14
86	35	13	12	198	108	57	61	3	32
72	14	38	29	91	28	39	49	87	3
45	85	23	98	65	84	26	71	32	59
44	85	32	96	53	48	51	76	87	12

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	1	1	0	0	0	0
0	0	0	1	1	1	1	1	0	0
0	0	0	1	1	1	1	1	0	0
0	0	0	0	1	1	1	1	0	0
0	0	0	0	1	1	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

Foreground

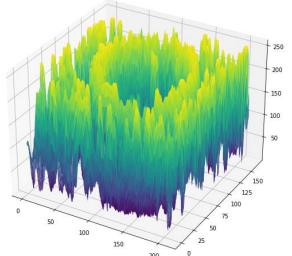
Background

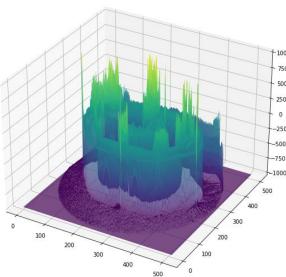


Segmentation Methods

Random Taxonomy

- Based on global knowledge:
 - Histogram-Based Thresholding
- Edge-based
 - Filters
- Region-Based
 - KNN/GMM
- A combination the two
 - Edge-based and Region-Based are duals for each other.







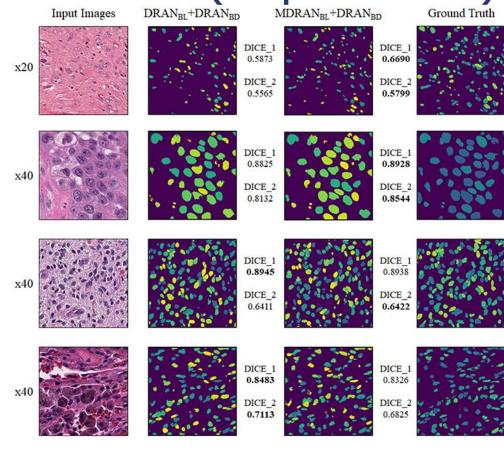
Clustering in Vision: Grouping Pixels

- Top down vs. bottom-up segmentation
- Top down: pixels belong together because they are from the same object
- Bottom up: pixels belong together because they look similar
- Hard to measure success:
 - Evaluation metrics depend on the application each measure a different aspect
 - <u>(55) Lena Maier-Hein @ ICBINB Seminar Series YouTube</u> Shows examples where the metrics are high, but the models still miss

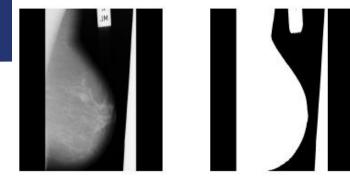


Segmentation via Classification (supervised)

- Classifier: KNN, Support Vector Machine (SVM), NN, etc.
- Features:
 - Voxel value
 - Voxel position
 - Gradient magnitude
 - Neighboring values
- Labels:
 - Foreground/Background (binary) or Class IDs (multiclass)
- Requires Labeled Training Data







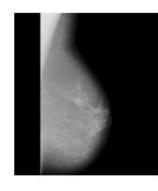
Segmentation via Clustering (Unsupervised)

K-Means

- Algorithm:
- Pick K feature space cluster centers at random
- While not converged
 - Assign each pixel to the nearest cluster
 - Recalculate cluster centers as the centroid of pixels in that cluster

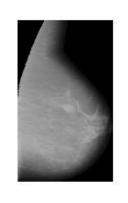
Fully Automatic Computer-Aided Detection of Breast Cancer based on Genetic Algorithm Optimization | IEEE Conference Publication | IEEE Xplore





(d)

(c)



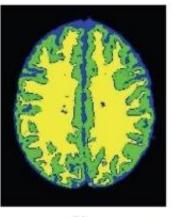
(e)

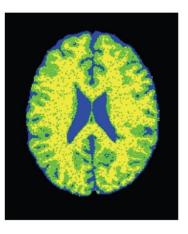


Segmentation via Clustering (Unsupervised)

Gaussian Mixture Models (GMM)

- Algorithm:
- Pick K feature space cluster centers using K-Means
- Initialize Gaussians (random mean & co-variance)
- While not converged
 - Calculate the probabilities for each pixel to each Gaussian
 - Recalculate the mean and the co-variance of each Gaussian





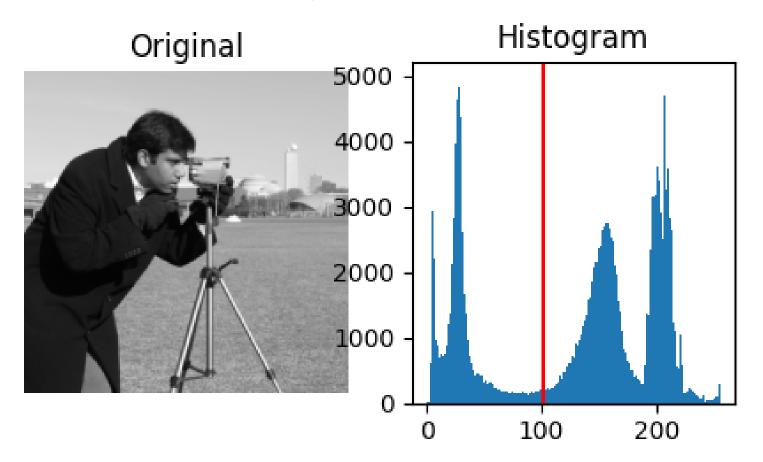
(c)



THRESHOLDING

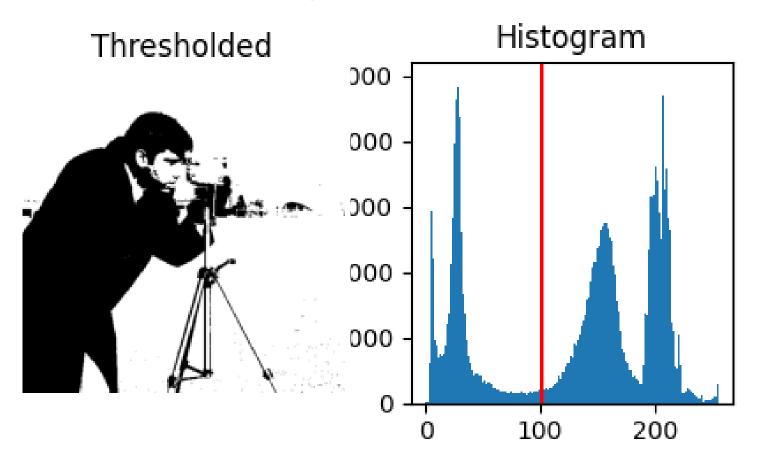


Thresholding





Thresholding





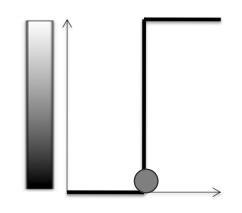
Intensity Thresholding

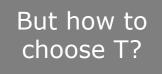
- Input: Gray-level images
- Output: binary Image
- Works best on high-contrast images

Exercise 1

Algorithm:

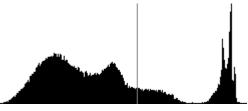
- Choose a threshold pixel value T
- For every pixel
 - if pixel ≥ T, label as foreground
 - Else: label as background













Choosing a Threshold Value

Example: Otsu's Method

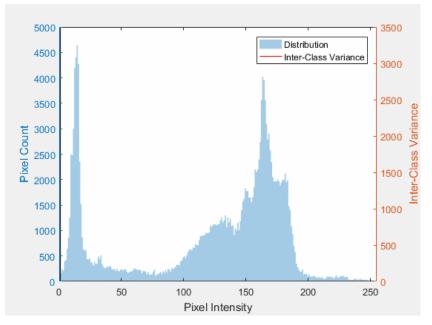
Searches the point *t* that minimizes the *variance* of foreground and background pixel values, weighted by class probabilities*.

$$\sigma_{\omega}^{2}(t) = P_{a}(t) \ \sigma_{a}^{2}(t) + P_{b}(t)\sigma_{b}^{2}(t)$$

Works well for bi-modal histograms.

* **Probabilities**: count how many pixels with those colors belong to the foreground, and how many to the background

Otsu Thresholding - The Lab Book Pages

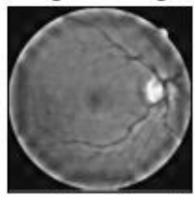




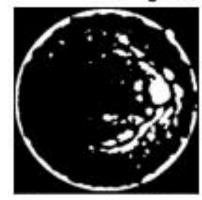
Adaptive Thresholding

- Global Thresholding:
 - operates on the whole image
- Adaptive Thresholding:
 - Using a running-window
 - Every patch is handled on its own

Original Image



Global Thresholding (v = 127)





Choosing a Threshold Value

- Many algorithms exist:
 - Maximum Entropy
 - Niblack
 - Li's
 - Kapur
 - Sauvola
 - Wolf ...
- Often, task-dependent
 - no single perfect threshold value exists.
- Thresholding leads to multiple disconnected components (i.e., "islands")

manuelaguadomtz/pythreshold: PyThreshold is a python package featuring Numpy/Scipy implementations of state-of-the-art image thresholding algorithms. (github.com)



Thresholding



What are your Take-Aways



Thresholding

Take-Aways

- A method to detect a middle point through the image histogram
- Clusters the image to foreground/background
- Operates either globally or on a running-window

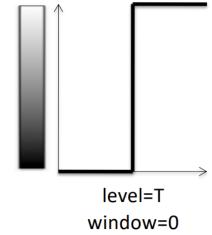


REGION-GROWING



Thresholding

Recap

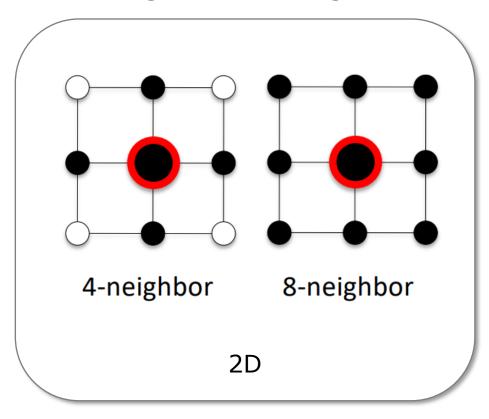


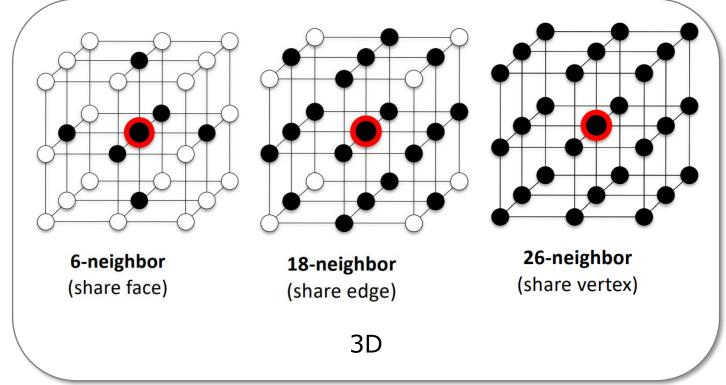
- Given a threshold level T, for every pixel:
 - If pixel ≥ T, label as foreground, else as background
- Otsu's Method
 - minimize variance of foreground and background pixel values weighted by class probabilities
- Maximum entropy
 - Maximize sum of each class's entropy: $H(t) = -\sum_i p(a_i) \log p(a_i) \sum_i p(b_i) \log p(b_i)$
- Adaptive/local Threshold
 - Running window (same as filters)
 - Local mean, local median, Sauvola, Perwit, ...



Connectivity

Defining Anatomic Regions Based on Contiguity







Region-Growing

Group pixels with similar properties – finding continuous regions

Algorithm:

- Select "seed" pixels according to some criteria
- Add to a region and push to the back of the queue
- While the queue is not empty
- For each neighbor of the front of the queue:
- If the neighbor meets the criteria and isn't in the region
- add to the region and push to the back of the queue
- Pop head of the queue

Example Implementation

Notes:

Criteria can be anything (e.g., global threshold)

* If you are familiar with Graph Theory: This is a breadth-first search

Iteration #1 with E=1221716

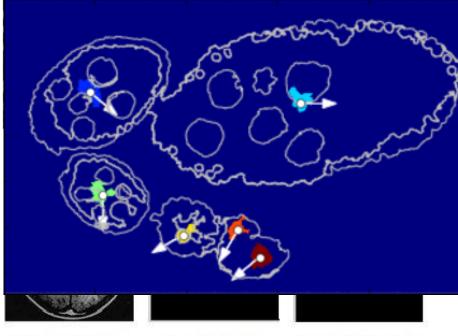


Figure 5(a). Original Image

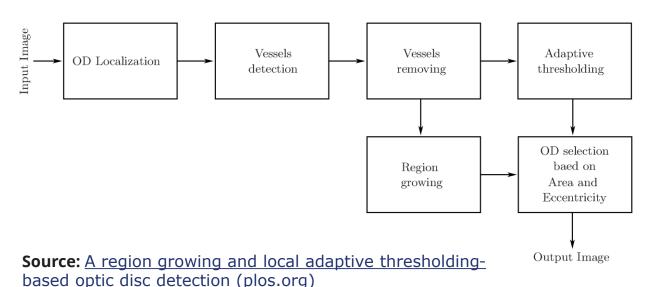
Figure 5(b). Seed points Figure 5(c). Result of Region growing

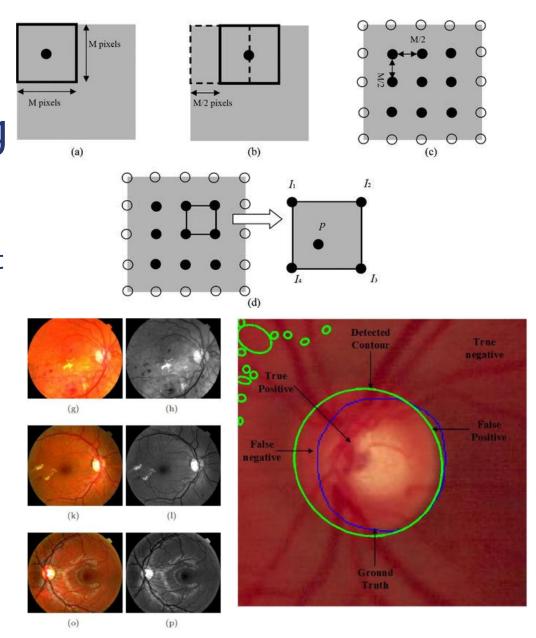
Figure 5. segmentation by Region Growing 22



Local Adaptive Thresholding

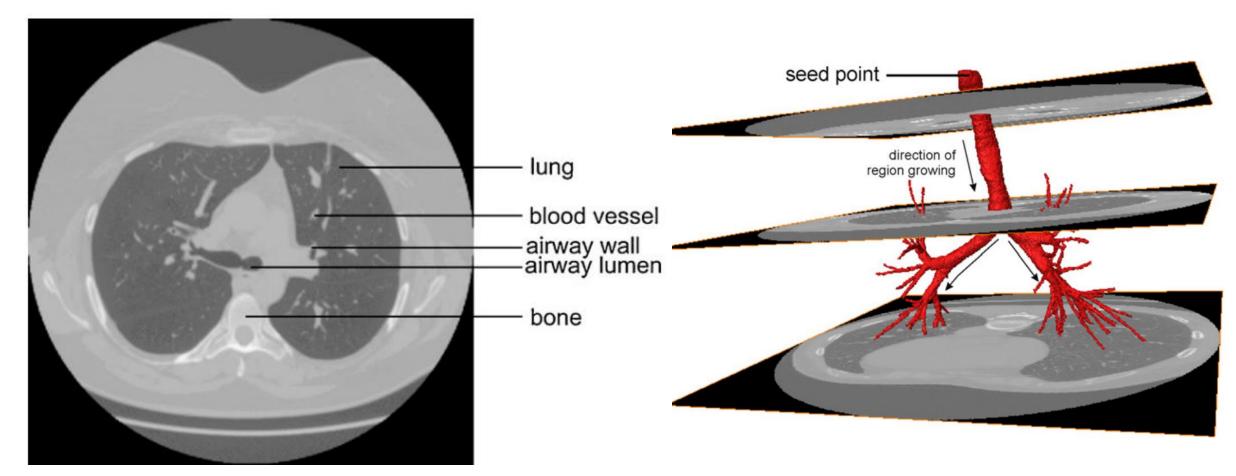
- Calculating the thresholds separately on different regions of the image.
- Note: Regions can be chosen 'smartly'







Region Growing – Best for a Continuous Region







Kernels & Filters

Kernel – a 2 parameters function: $f(x_1, x_2)$

These parameters can be scalars or matrices

Q: Where have we met (and used) **Kernels** before in ML?



Image Kernels (Filters)

- Kernel: a fixed matrix of numbers
- Calculates an average on the image, using a sliding window

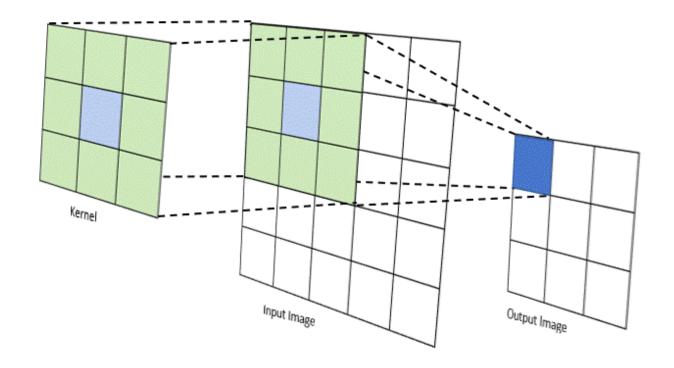
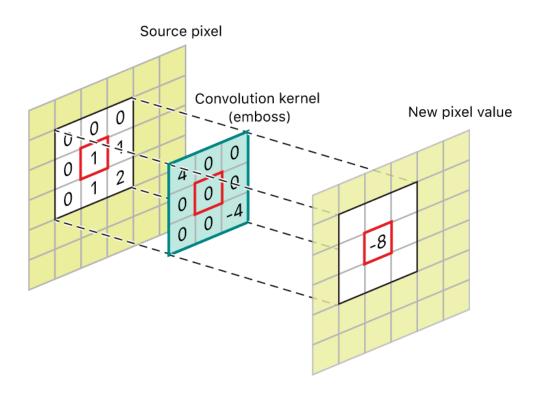




Image Kernels (Filters)



- Different Kernels exist.
 - Used for: blur, sharpen, shift
- It can also detect **edges**, and more.
- Try it yourself: https://setosa.io/ev/image-kernels/

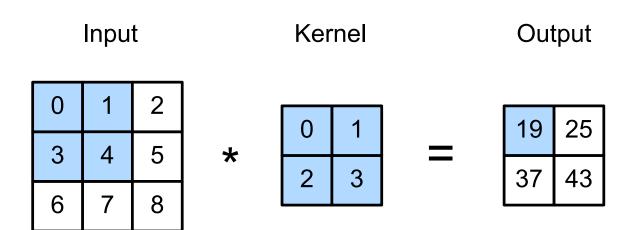
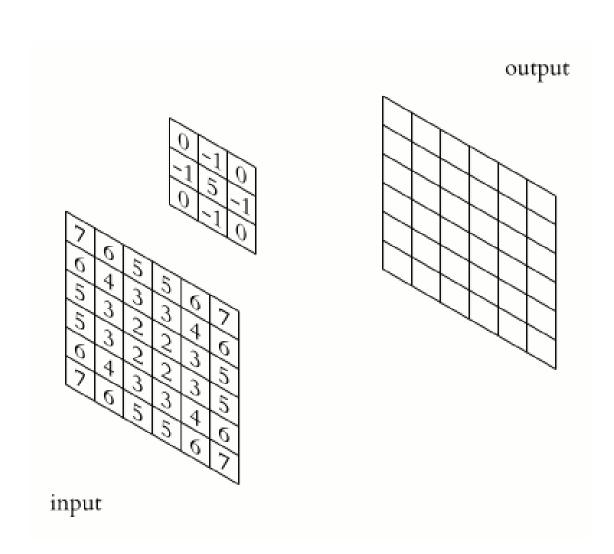


Image Kernels explained visually (setosa.io)



Aka "Convolution"

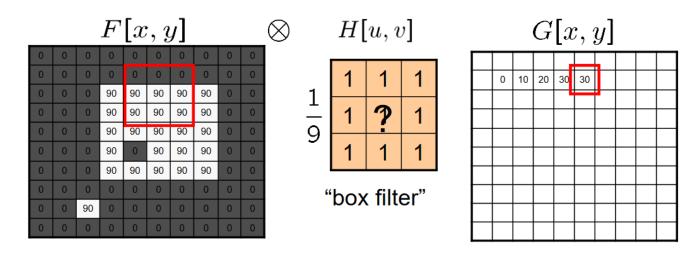
(we'll meet it again in neural networks)





Averaging Filter

What values belong in the kernel H for the moving average example?



$$G = H \otimes F$$

Removes high-frequency components from the image ("low-pass filter").

Gaussian Filters

• What if we want nearest neighboring pixels to have the most influence on the

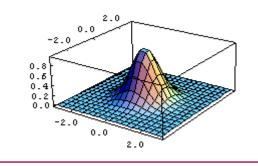
output?

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

$$\overline{F[x,y]}$$

This kernel approximates a 2d Gaussian function:

$$h(u,v) = \frac{1}{2\pi\sigma^2} e^{-\frac{u^2+v^2}{\sigma^2}}$$

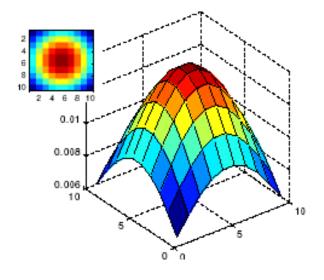




Gaussian Filters

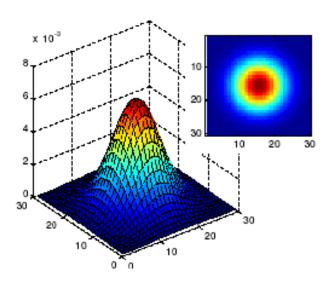
Hyperparameters affect the result: Kernel Size

Note: Gaussian function has infinite support, but discrete filters use *finite* kernels



 $\sigma = 5$ with 10×10 kernel

Kernel Size - determines the range the smoothing takes into account



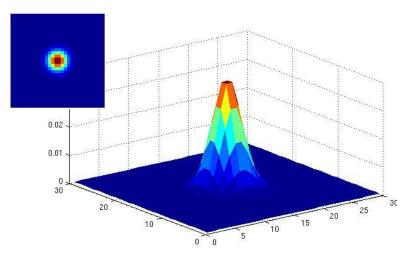
 $\sigma = 5$ with 30×30 kernel

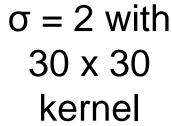


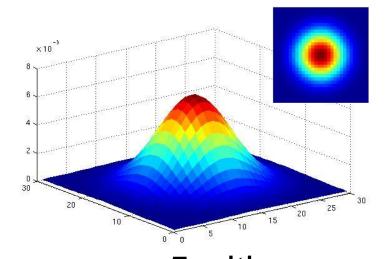
Gaussian Filters

Hyperparameters affect the result: Gaussian Variance

Variance of Gaussian - determines extent of smoothing







$$\sigma = 5$$
 with 30×30 kernel



Gaussian Filter

