Intensity-based Registration



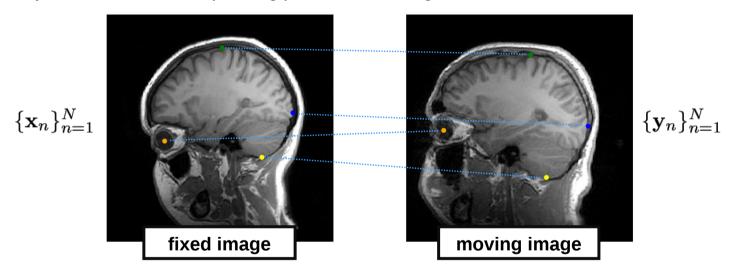
Medical Image Analysis

Koen Van Leemput

Fall 2023

Recall landmark-based registration

ightharpoonup Manually annotate N corresponding points in two images:



Register the images by minimizing the distance between matching point pairs:



$$E(\mathbf{w}) = \sum_{n=1}^{N} \|\mathbf{y}_n - \mathbf{y}(\mathbf{x}_n, \mathbf{w})\|^2$$

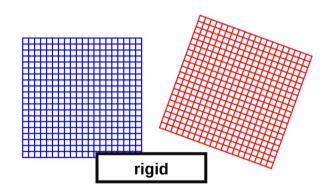
Spatial transformation model

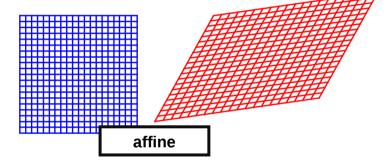
Spatial transformation models

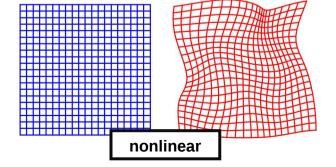
Spatial transformation y(x, w):

 \checkmark maps world positions \mathbf{x} in the fixed image to world positions \mathbf{y} in the moving image

controlled by parameters w

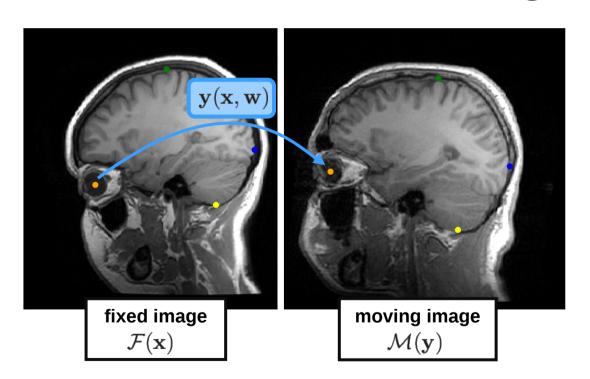


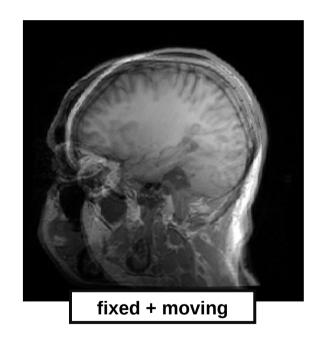






Landmark-based registration

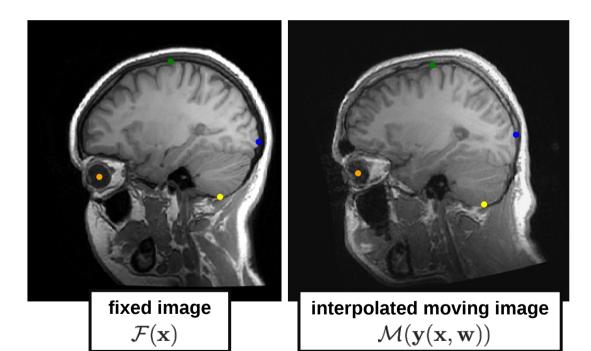


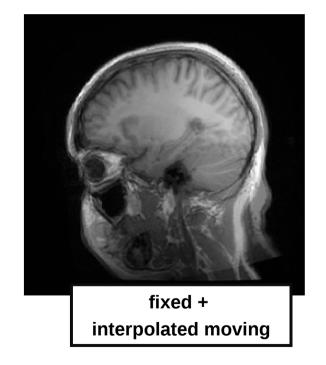




Before registration

Landmark-based registration

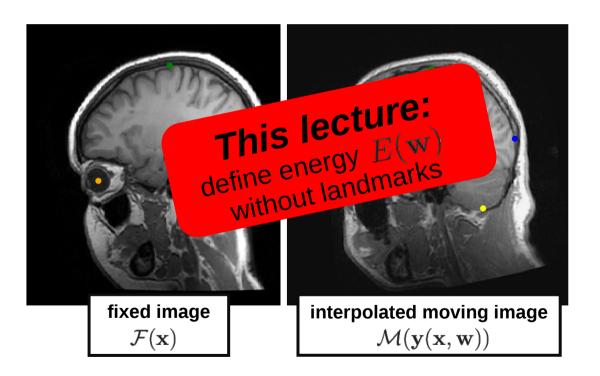


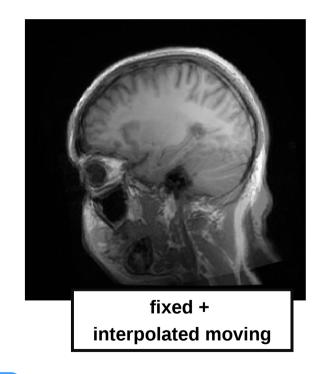




After registration

Landmark-based registration

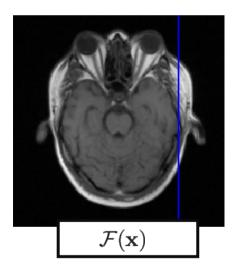


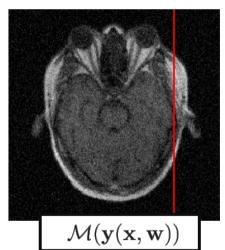


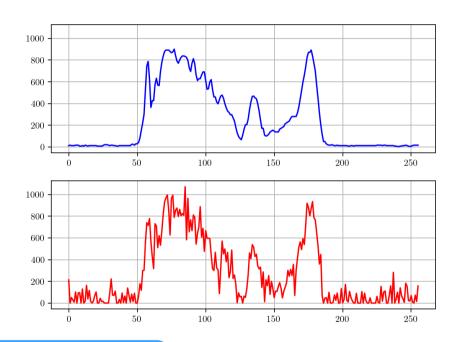


After registration

Images have similar intensity characteristics

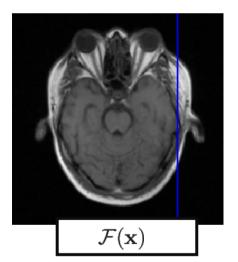


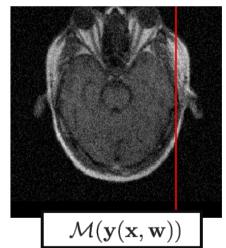


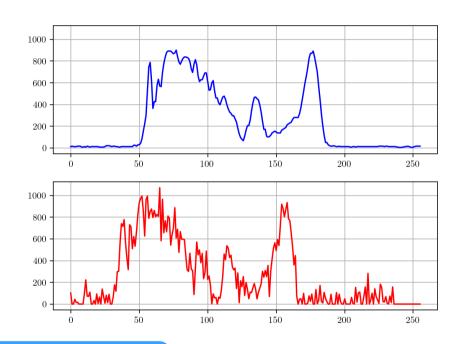




Images have similar intensity characteristics

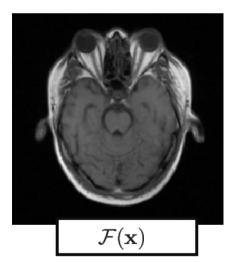


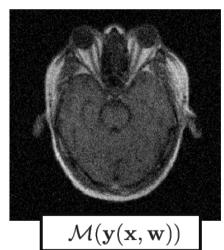


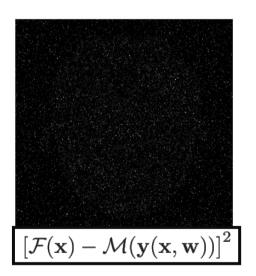




Images have similar intensity characteristics





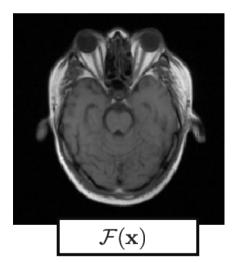


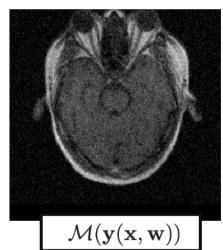
$$E(\mathbf{w}) = \sum_{n=1}^{N} \left[\mathcal{F}(\mathbf{x}_n) - \mathcal{M}(\mathbf{y}(\mathbf{x}_n, \mathbf{w})) \right]^2$$

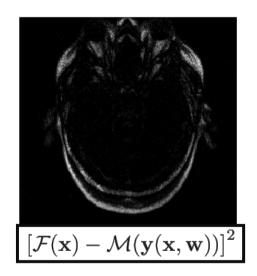
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Aalto-universitetet
Aalto University

sum over all voxels

Images have similar intensity characteristics

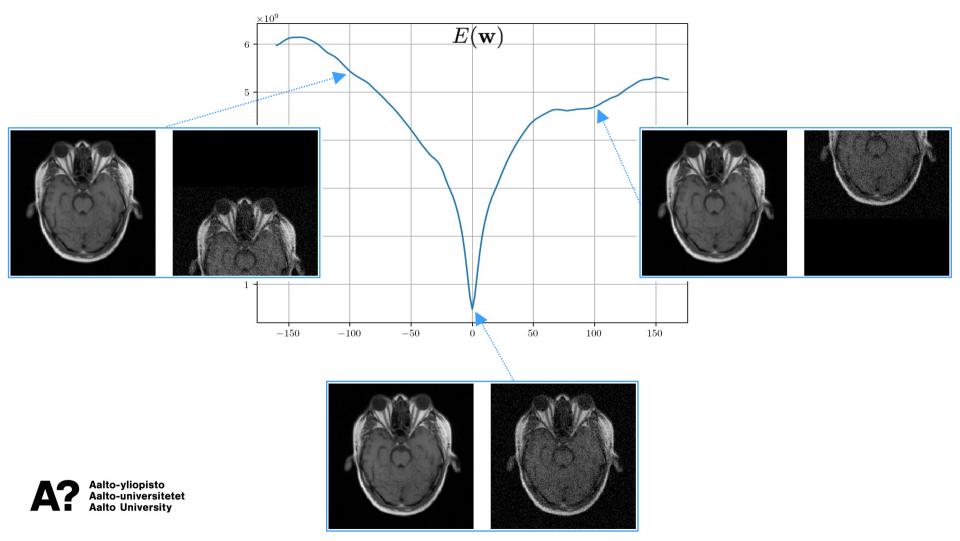




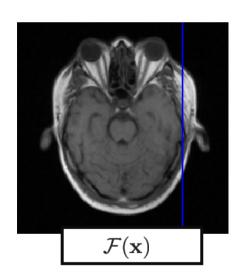


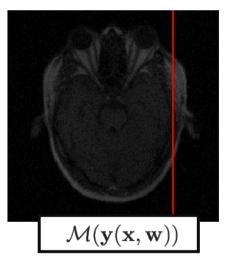


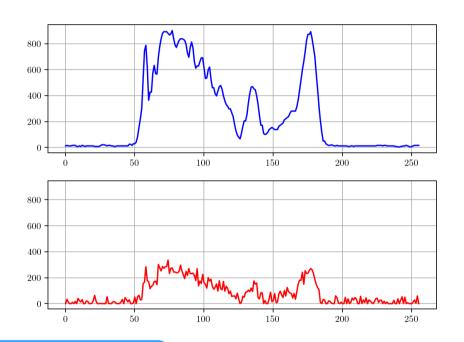
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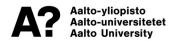


Same but images are scaled differently

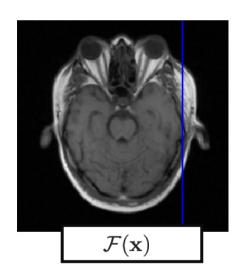


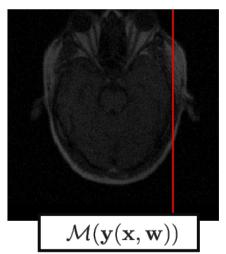


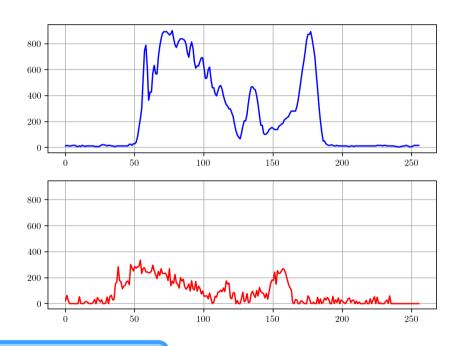


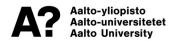


Same but images are scaled differently







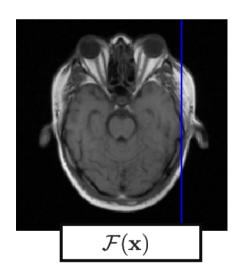


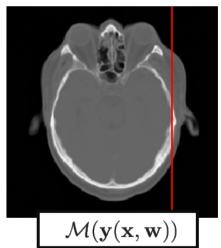
Same but images are scaled differently

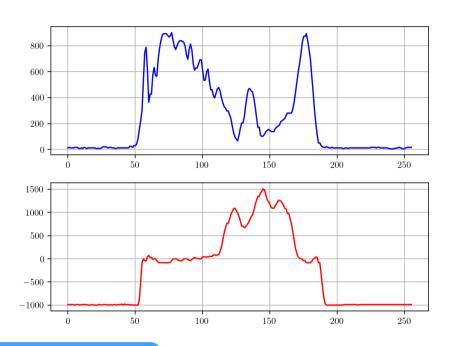
TODO: robust min/max histogram from cumulative histogram



Images have different intensity characteristics

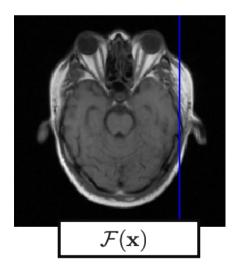


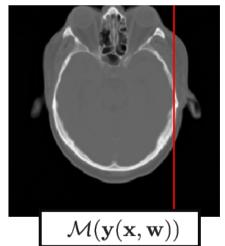


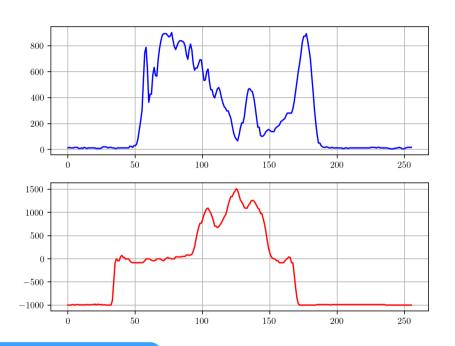




Images have different intensity characteristics

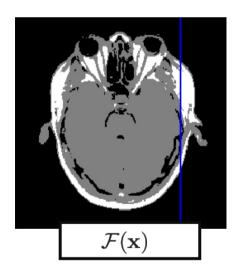


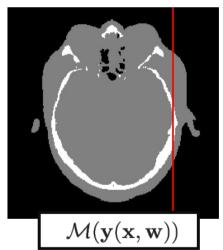


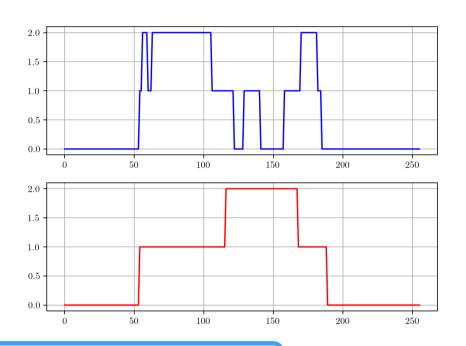


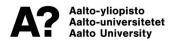


Images have different intensity characteristics



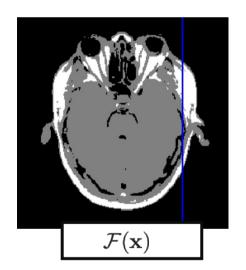


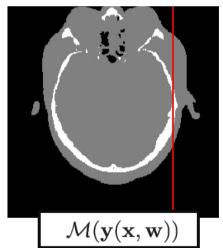


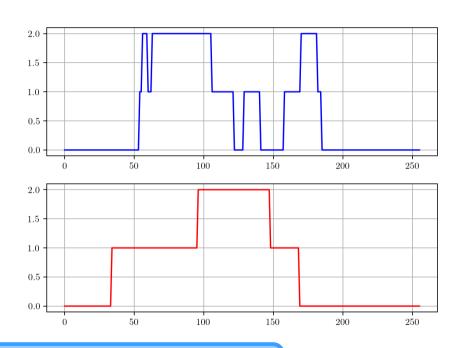


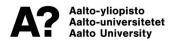
Easier task: what's a good energy function $E(\mathbf{w})$ now?

Images have different intensity characteristics

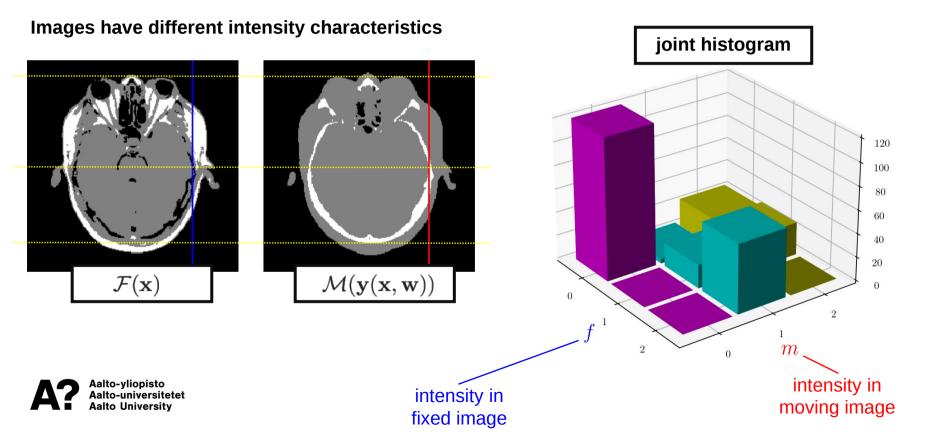


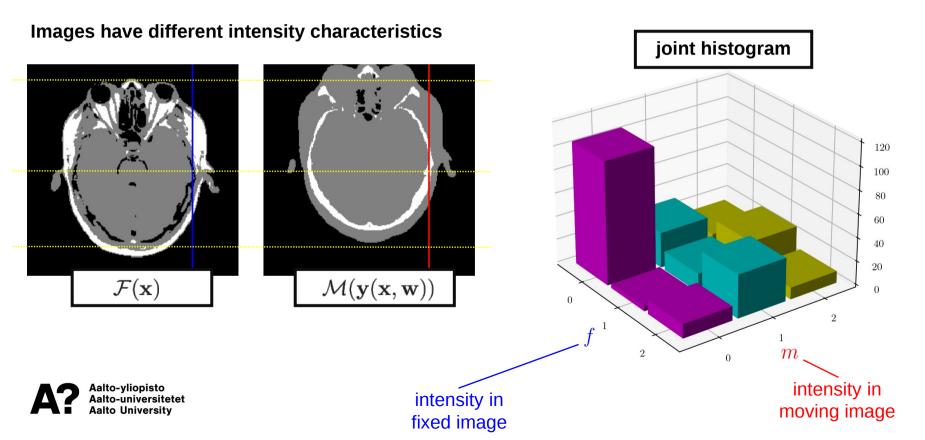




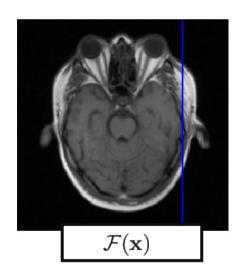


Easier task: what's a good energy function $E(\mathbf{w})$ now?

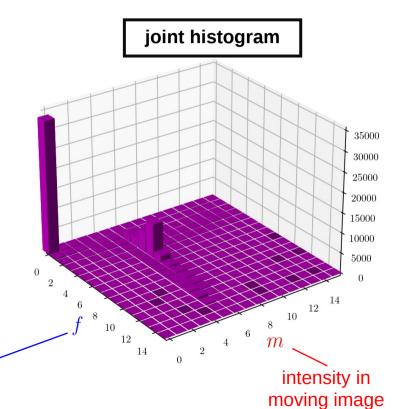




Images have different intensity characteristics



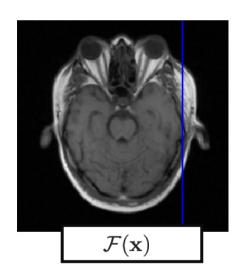


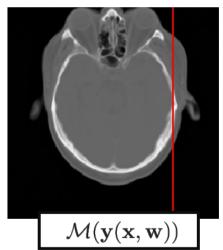


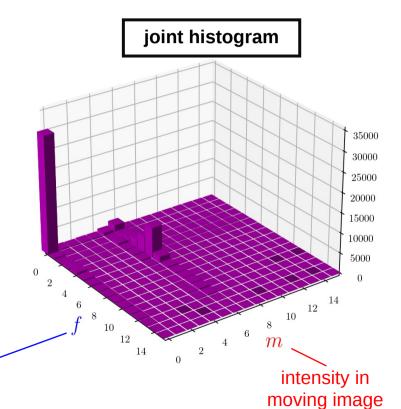


intensity in fixed image

Images have different intensity characteristics









intensity in fixed image

A bit of information theory...

Imagine that a coin is "rigged":

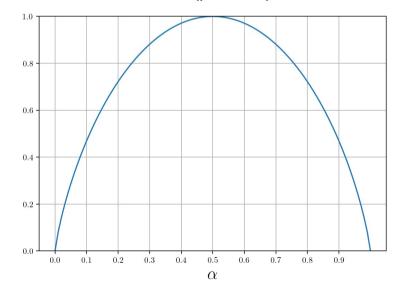
- $m \prime$ lands on heads with probability $0 \le lpha \le 1$

"heads"

✓ The minimum number of bits required to store/communicate this result is (per toss):

$$-\alpha \log_2(\alpha) - (1 - \alpha) \log_2(1 - \alpha)$$

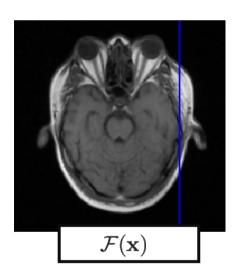
"entropy"

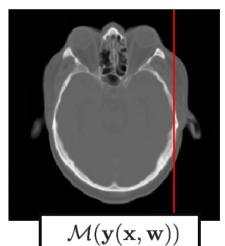


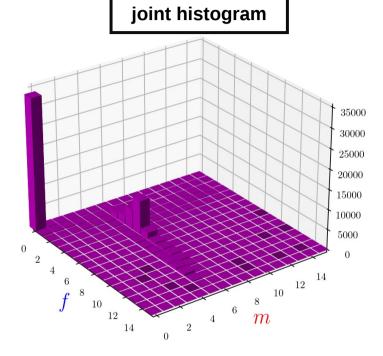
"tails"



Images have different intensity characteristics





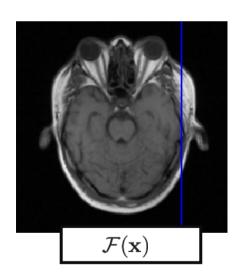


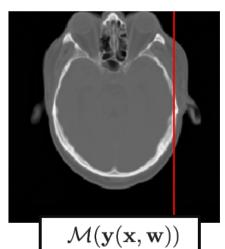


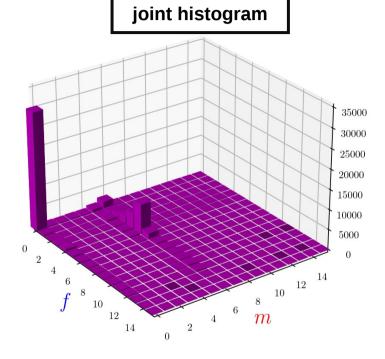
$$E(\mathbf{w}) = H_{F,M}$$
 where $H_{F,M} = -\sum_{f=1}^B \sum_{m=1}^B p_{f,m} \log(p_{f,m})$

__ normalized histogram counts

Images have different intensity characteristics

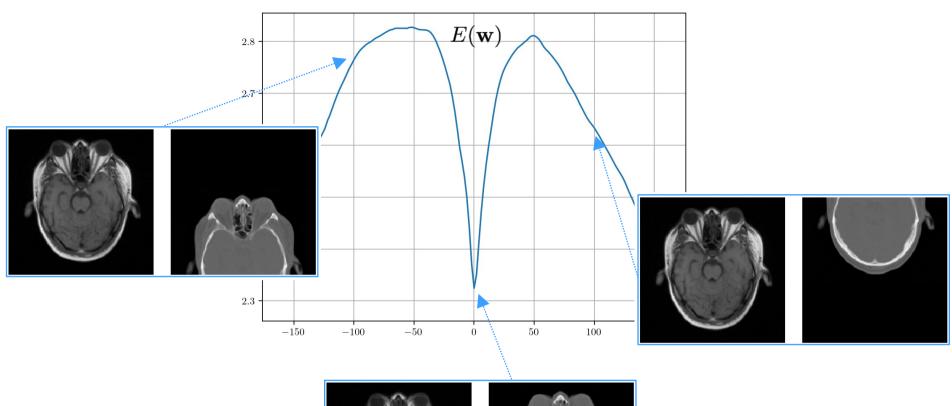




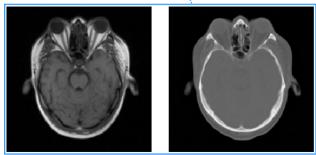




$$E(\mathbf{w}) = H_{F,M}$$
 where $H_{F,M} = -\sum_{f=1}^B \sum_{m=1}^B p_{f,m} \log(p_{f,m})$

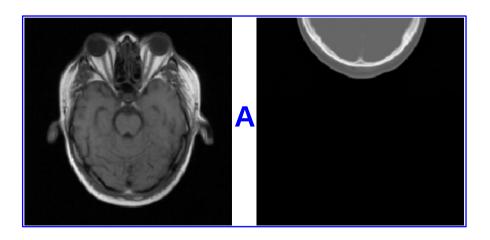


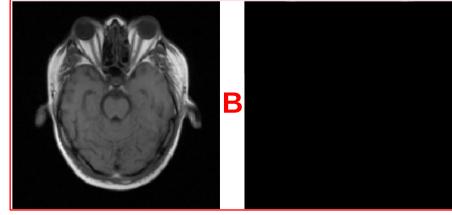




Diagnosing the problem

Question: which image pair takes more bits to encode?

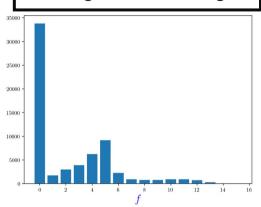






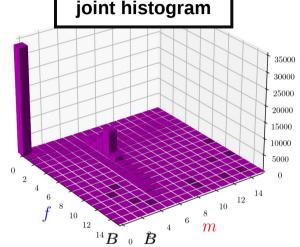
Solution

histogram fixed image



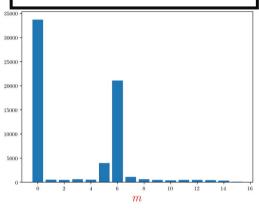
$$H_F = -\sum_{f=1}^B p_f \log(p_f)$$

joint histogram



$$H_F = -\sum_{f=1}^{D} p_f \log(p_f)$$
 $H_{F,M} = -\sum_{f=1}^{D} \sum_{m=1}^{D} p_{f,m} \log(p_{f,m})$ $H_M = -\sum_{m=1}^{D} p_m \log(p_m)$

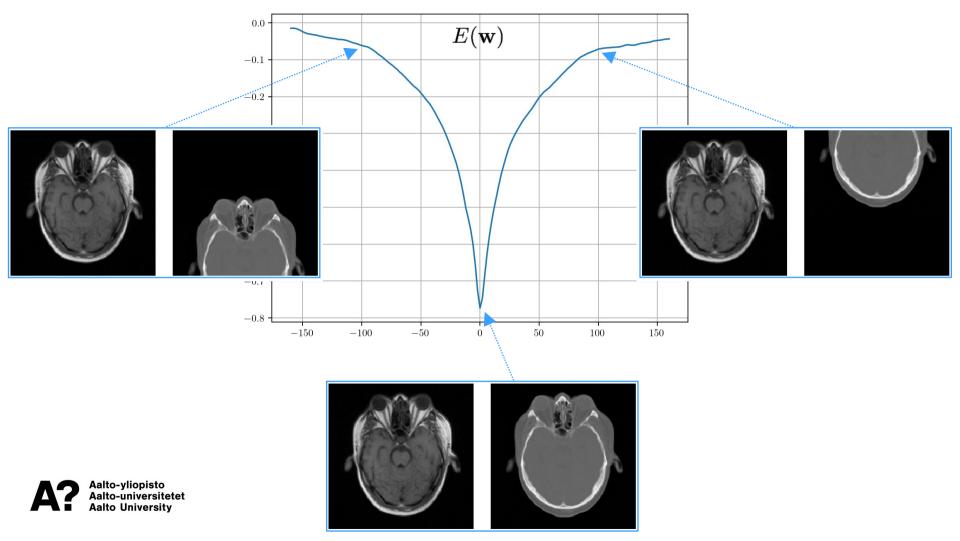
histogram moving image



$$H_M = -\sum_{m=1}^B p_m \log(p_m)$$



 $E(\mathbf{w}) = H_{F,M} - H_F - H_M$ (negative "mutual information")



Numerical optimization

Find transformation parameters \mathbf{w} that minimize $E(\mathbf{w})$

