

# Medical Image Processing for Diagnostic Applications

## Filtering in Spatial Domain

Online Course – Unit 23

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Pattern Recognition Lab (CS 5)



# Topics

Homomorphic Unsharp Masking

Polynomial Surface Fitting

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# Homomorphic Unsharp Masking

**Apply mean normalization.**

# Homomorphic Unsharp Masking

Homomorphic Unsharp Masking (HUM) is one of the simpler IIH correction methods, and also the most commonly used approach. HUM requires the computation of:

- the global mean value  $\mu$  of the intensity distorted image,
- local mean values  $\mu_{i,j}$  evaluated in a neighborhood of each pixel  $(i,j)$ .

If the multiplicative model is used, the estimated intensity corrected value  $f_{i,j}$  is then computed pixelwise in the following manner:

$$f_{i,j} = \frac{\mu}{\mu_{i,j}} g_{i,j}.$$

g is the already filtered image ?  
the one where we lost the mean value because we used a bandwidth filter ?

# Homomorphic Unsharp Masking

A few remarks on homomorphic unsharp masking:

- This IIH-correction method relies on the assumption that the local means in an image are equal to the global mean in the absence of IIH.
- Differences between the global and local means are thus caused by the bias field only.
- This assumption only holds if the neighborhood used for computing the local mean contains a representative sample of the tissue types in the image.
- The size of the neighborhood has to be chosen carefully which is an experimental problem.  
size might depend on body region

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# Polynomial Surface Fitting

**Approximate the low frequency bias field in the additive model by a multivariate regression polynomial and eliminate the bias by subtraction.**

biast fiel -> low order polinomial

# Polynomial Surface Fitting

## The basic idea:

1. The logarithmized image is considered as a 2-D function, where the pixel coordinates  $(i, j)$  denote the sampling points and the intensities  $\log g_{i,j}$  denote the associated function values.
2. Fit a parametric, smooth surface to the logarithm of the intensity values.
3. Estimate the parameters by minimizing the sum of **squared differences** of the surface points and the logarithmized image intensities.
4. The resulting surface is then subtracted from the logarithmic image.



# Polynomial Surface Fitting

Recalling our approach for spatial distortion correction, we define now a parametric mapping for intensity undistortion:

- we consider the image point at  $(i, j)$ ,
- we assume separable base functions,
- and thus require univariate base functions  $b_k : \mathbb{R} \rightarrow \mathbb{R}$ ,  $k = 0, \dots, d$ , and
- the coefficients of the polynomials  $b_k$  in  $i$  and  $j$  are denoted by  $u_{k,l} \in \mathbb{R}$ .

Accordingly, the polynomial that approximates the bias field is defined by:

$$g_{i,j} \approx \sum_{k=0}^d \sum_{l=0}^{d-k} u_{k,l} b_k(i) b_l(j).$$

# Polynomial Surface Fitting

here  $g$  is the value we get directly from the scanner

The resulting least square estimation problem is:

$$[\hat{u}_{k,l}] = \arg \min_{u_{k,l}} \sum_{i,j=0}^{N-1} \left\| g_{i,j} - \sum_{k=0}^d \sum_{l=0}^{d-k} u_{k,l} b_k(i) b_l(j) \right\|^2.$$

This optimization problem can be solved by computing the **SVD** of the associated measurement matrix.

The final bias field estimate is:

$$b_{i,j} = \sum_{k=0}^d \sum_{l=0}^{d-k} \hat{u}_{k,l} b_k(i) b_l(j).$$

**Exercise:** Compute the measurement matrix for  $b_k(i) = i^k$  and  $b_l(j) = j^l$ , and think about a proper scaling.

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# Take Home Messages

two different methodes ?!!

- Spatial filtering can also be used for bias field correction. In homomorphic unsharp masking the **local means** are **adapted to** the **global mean**.
- Similar to the undistortion algorithms from earlier units, a **polynomial** can be used to estimate and correct the bias field.

## Further Readings

The webpage of the [National High Magnetic Field Laboratory](#) can be one starting point for more detailed information regarding MRI. For an initial overview of the technology, the following article is worth reading:

[MRI: A Guided Tour](#) by Kristen Coyne.

If you want to know more about segmentation of MR images, e. g., consult the [Google Scholar record](#) of ‘Sandy’ Wells’ publications.

Another article worth reading is this survey paper on algorithms for intensity correction methods:

[Zujun Hou](#). “A Review on MR Image Intensity Inhomogeneity Correction”. In: *International Journal of Biomedical Imaging* 2006. Article ID 49515 (Feb. 2006), pp. 1–11. DOI: 10.1155/IJBI/2006/49515