Simultaneous Reconstruction of Attenuation and Activity for non-TOF PET/MR Using MR Prior Information

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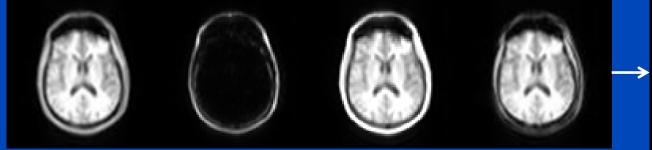
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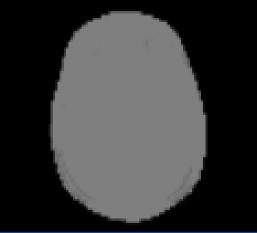
Introduction Motivation

- Quantitative PET imaging requires accurate attenuation correction (AC).
- Standard MR-based attenuation correction (MRAC) neglects bone attenuation.

MR images obtained using two-point Dixon VIBE sequence



Attenuation Map



Neglecting bone yields activity underestimation values of up to 30%^{1.}

[1] Samarin *et al.*, "PET/MR imaging of bone lesions - implications for PET quantification from imperfect attenuation correction," *Eur. J. Nucl. Med. Mol. Imaging*, 39(7), 1154–60, 2012.

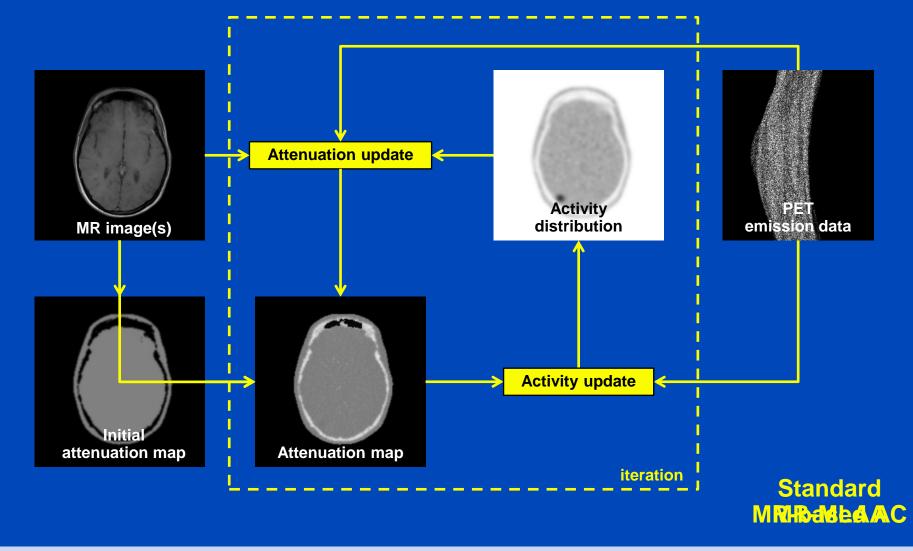


Introduction

- Improve AC for non-TOF PET/MR by simultaneous reconstruction of attenuation and activity distributions from PET emission data using MR prior information.
- The presented algorithm is an extension of the maximum-likelihood reconstruction of attenuation and activity (MLAA)¹ for non-TOF PET/MR, called MR-MLAA.



Algorithm Workflow





Algorithm Cost Function

Cost function C

 $C(\boldsymbol{\lambda}, \boldsymbol{\mu}) = L(\boldsymbol{\lambda}, \boldsymbol{\mu}) + L_{\mathrm{S}}(\boldsymbol{\mu}) + L_{\mathrm{I}}(\boldsymbol{\mu})$

- Log-likelihood *L* $L(\lambda, \mu) = \sum_{j} (p_j \ln \hat{p}_j - \hat{p}_j)$ with $\hat{p_j} = e^{-\sum_i \mu_i l_{ij}} \sum_i \lambda_i M_{ij}$
- Smoothing prior L_S
- Intensity prior L

 λ Activity μ Attenuation

 p_j Measured projections along LOR j

 l_{ij} Intersection length of voxel *i* and LOR *j*

 ${\cal M}\,$ System matrix



Algorithm Update Equations

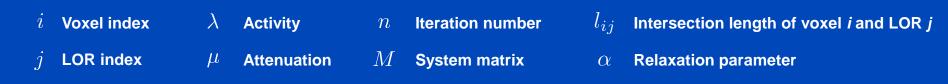
Activity Update (AW-MLEM)

$$\lambda_{i}^{(n+1)} = \lambda_{i}^{(n)} \frac{1}{\sum_{j} M_{ij} a_{j}^{(n)}} \sum_{j} M_{ij} \frac{p_{j}}{\sum_{i} \lambda_{i}^{(n)} M_{ij}}$$

Attenuation Update¹

$$\mu_j^{(n+1)} = \mu_j^{(n)} + \alpha \frac{\sum_j \left(M_{ij} \left(a_j^{(n)} \sum_i \lambda_i^{(n)} M_{ij} - p_j \right) \right) + \frac{\partial}{\partial \mu_i} (L_{\mathrm{S}} + L_{\mathrm{I}})}{\sum_j \left(M_{ij} a_j^{(n)} \sum_i \lambda_i^{(n)} M_{ij} \sum_i l_{ij} \right) - \frac{\partial^2}{\partial \mu_i^2} (L_{\mathrm{S}} + L_{\mathrm{I}})}$$

with attenuation factor
$$a_{j}^{(n)} = e^{-\sum_{i} \mu_{i}^{(n)} l_{ij}}$$
 Prior



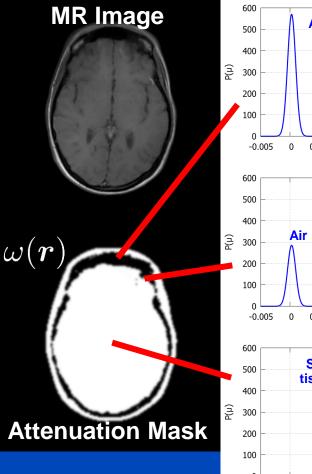
[1] Nuyts et al., "Iterative reconstruction for helical CT: a simulation study," Phys. Med. Biol., 43(4), 729–737, 1998.

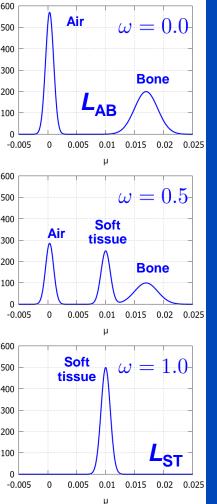
Algorithm **Prior Information**

- Optimizing the cost function $C(\lambda, \mu)$ with both the activity and attenuation distribution unknown is an ill-conditioned problem.
- Prior information can help to drive the algorithm towards a 'desired' solution.
- Smoothing prior L_S
 - Favors smooth attenuation map
 - Defined as logarithm of a Gibbs probability distribution
- Intensity prior L_I
 - Voxel-dependent Gaussian-like probability distribution of predefined attenuation coefficients (e.g., for soft tissue, air, bone)
 - Voxel-dependency is based on the MR information



Algorithm Intensity Prior I





- Use the MR image to create a mask defining air/bone and soft tissue
- Smooth mask
- Define intensity prior $L_{\rm I}$ as linear combination of air/bone intensity prior $L_{\rm AB}$ and soft tissue intensity prior $L_{\rm ST}$:

$$L_{\mathrm{I}} = (1 - \omega)\beta_{\mathrm{AB}}L_{\mathrm{AB}} + \omega\beta_{\mathrm{ST}}L_{\mathrm{ST}}$$

 $\omega({m r})$

Voxel-dependent weighting factor, given by the attenuation mask

 $eta_{ ext{AB}}, \, \overline{eta_{ ext{ST}}}$ Global weighting factors

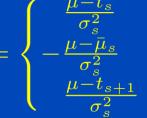


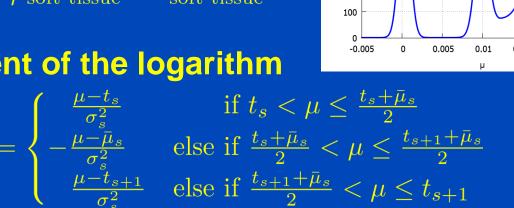
Algorithm **Intensity Prior II**

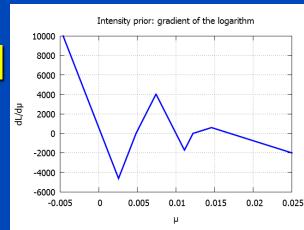
- Express the priors L_{AB} and L_{ST} as a combination of Gaussian-like functions and choose S pre-defined attenuation coefficients for each prior.
 - Air/Bone: $S = 2; \ \bar{\mu}_{air} \pm \sigma_{air}, \ \bar{\mu}_{bone} \pm \sigma_{bone}$
 - Soft tissue: S = 1; $\bar{\mu}_{\text{soft tissue}} \pm \sigma_{\text{soft tissue}}$

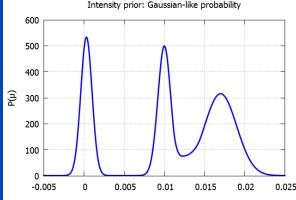
Define the gradient of the logarithm •

$$\frac{\partial L_{\rm AB/ST}(\mu,\bar{\mu}_s;s=0,\cdots,S-1)}{\partial\mu} =$$







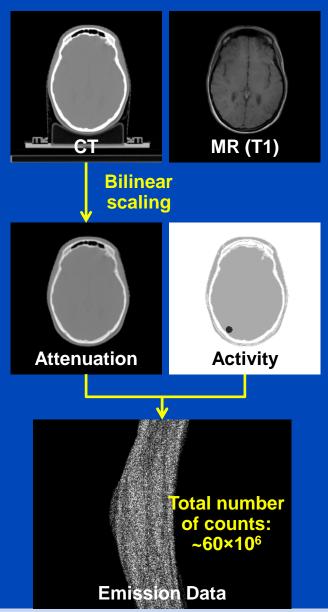


 $\bar{\mu}_{s-1} < t_s < \bar{\mu}_s$ Intersection points t_s between neighboring Gaussians



Experiments Data

- Use pair of co-registered MR/CT head patient data
- Simulate 3D activity distribution:
 - Fat/Soft tissue: 5-10 kBq/mL
 - Air/bone: 0 kBq/mL
 - Lesions: 25 kBq/mL
- Simulate 3D PET emission data accounting for
 - Poisson noise
 - attenuation
- Use Siemens Biograph mMR Geometry





Experiments Reconstructions

Ground Truth

AW-OSEM reconstructions using

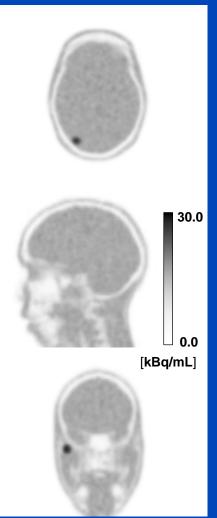
- the true attenuation for AC (ground truth)
- standard MR-based AC (MRAC)
- the final attenuation map from MR-MLAA for AC
 - » Mask derived from MR image (standard)
 - » Mask derived from attenuation image (idealized)

Reconstruction parameters:

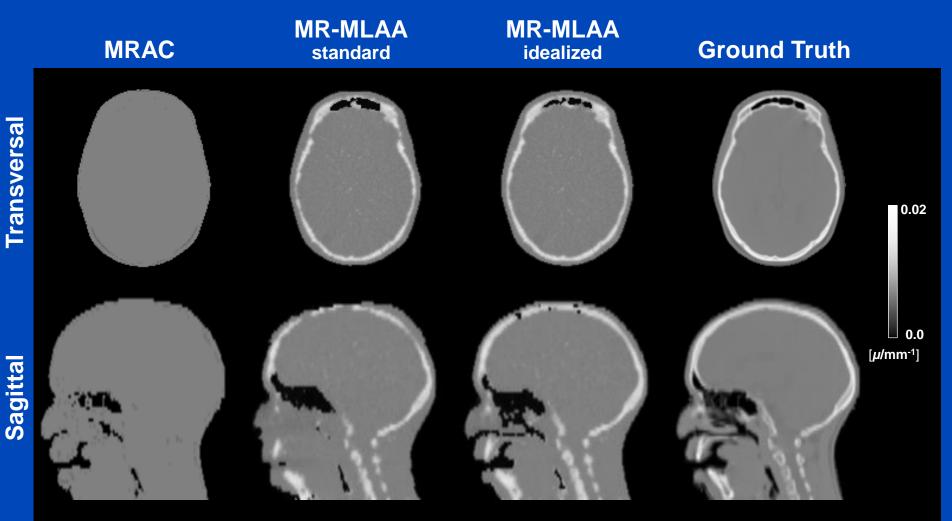
- Iterations: 3
- Subsets: 21
- Gaussian post-smoothing: FWHM = 5 mm

Quantitative Evaluation

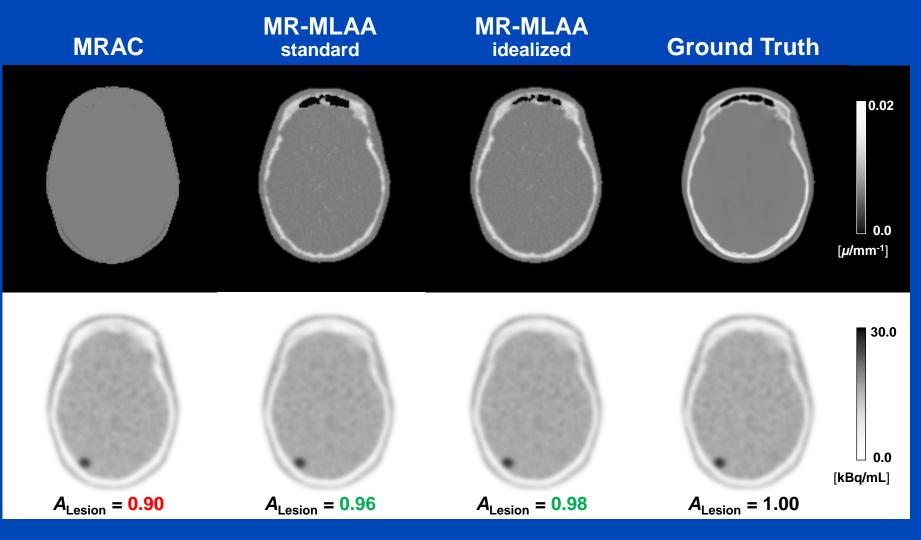
- Measure relative mean activity in ROIs corresponding to simulated lesions
- Calculate activity difference to ground truth



Results Attenuation



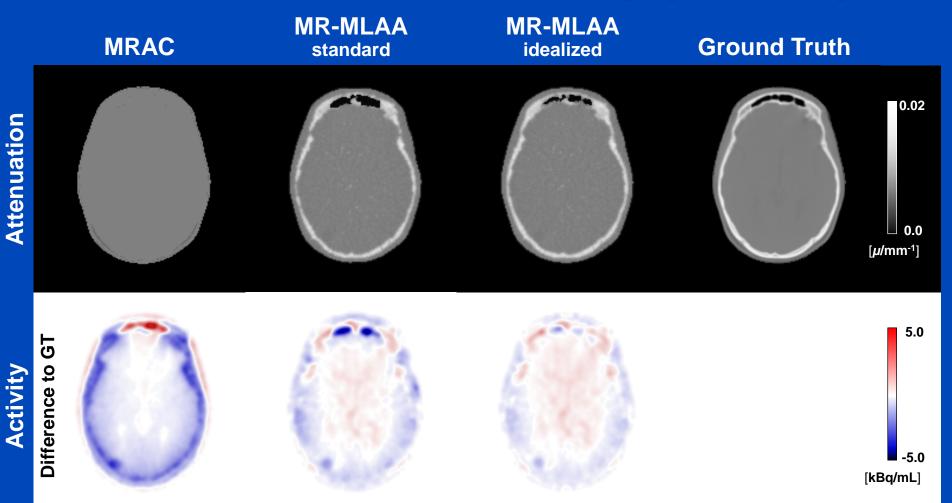




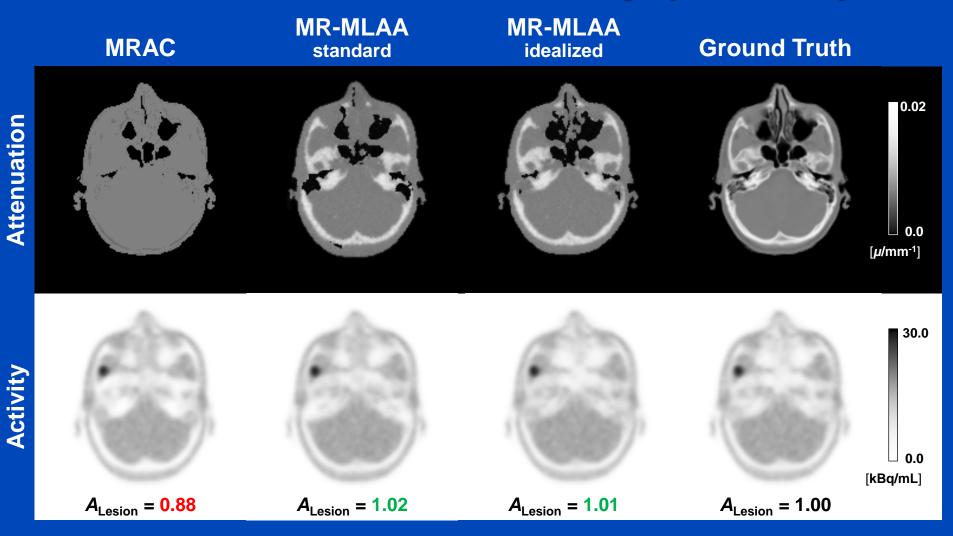
Attenuation

Activity

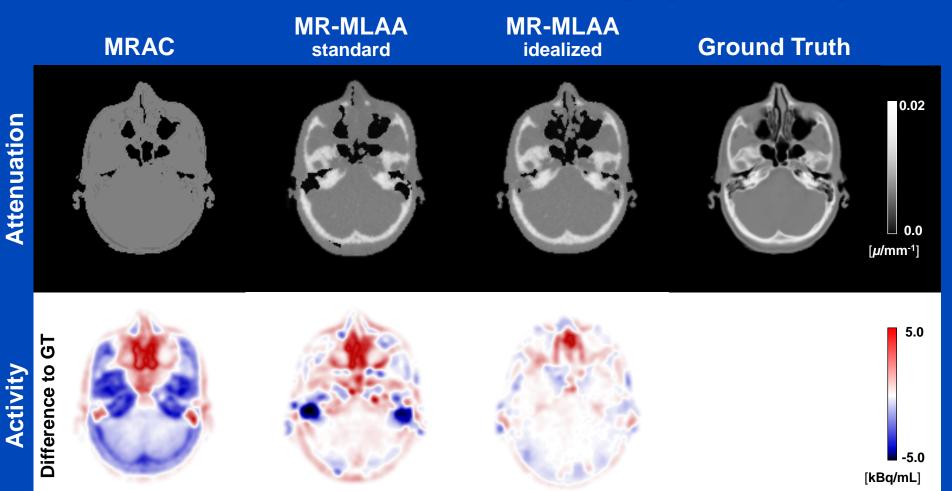














Conclusion

- MR-MLAA significantly improves bone attenuation estimation compared to MRAC.
- MR-MLAA improves PET quantification compared to MRAC, especially for regions close to bone tissue.
- Challenges
 - Small air cavities (e.g., nasal sinuses)
 - Misclassification of air as bone or soft tissue leads to increased activity values
 - Thin bone structures



Outlook

- Clinical data coming soon.
- Potential improvements
 - Sophisticated segmentation technique to create attenuation mask from MR image(s) (e.g., based on UTE images)
 - Additional prior information from non attenuation-corrected (NAC) images
 - Time-of-flight (TOF) information
- Adaption to whole-body PET/MR requires
 - additional tissue classes (e.g., fat)
 - proper handling of truncated MR data



Thank You!

The 4th International Conference on Image Formation in X-Ray Computed Tomography

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Conference Chair Marc Kachelrieß, German Cancer Research Center (DKFZ), Heidelberg, Germany

This presentation will soon be available at www.dkfz.de/ct.

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