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Feasibility of basis material decomposition with multilayer detectors

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Outline

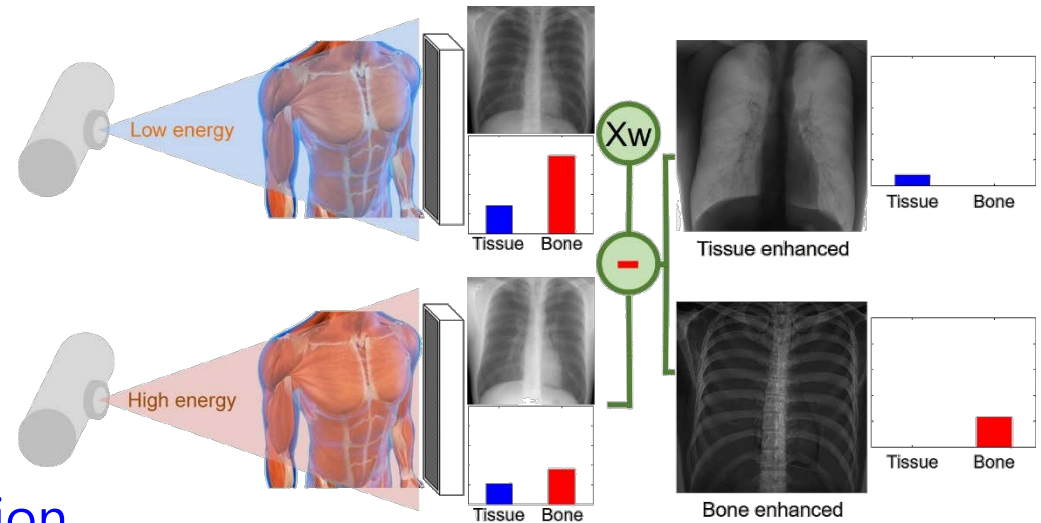
- Introduction
- Theoretical background
 - Basis function
 - Material decomposition algorithm
- Materials & Methods
- Results
- Conclusion

Dual-energy imaging

- The dual-energy (DE) imaging increases the **conspicuity** and allows a better view of the lesion to be seen

- Two approaches

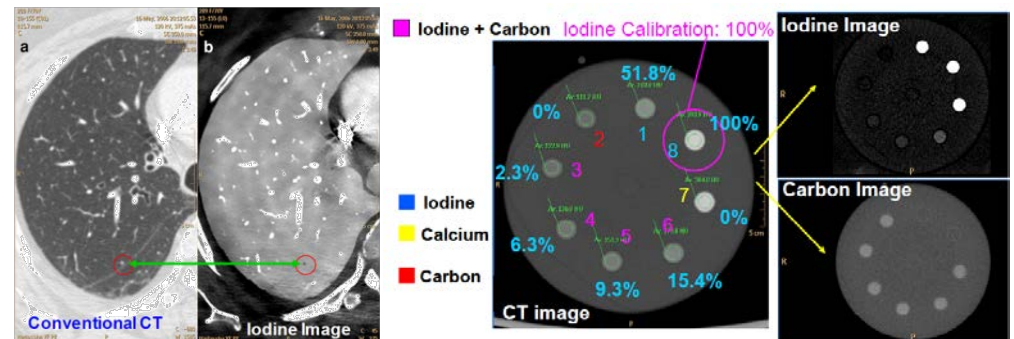
- Energy subtraction
weighted subtraction of images taken at two different energies



- Basis material decomposition

decomposition of the measured data or images into contributions due to the two "basis materials"

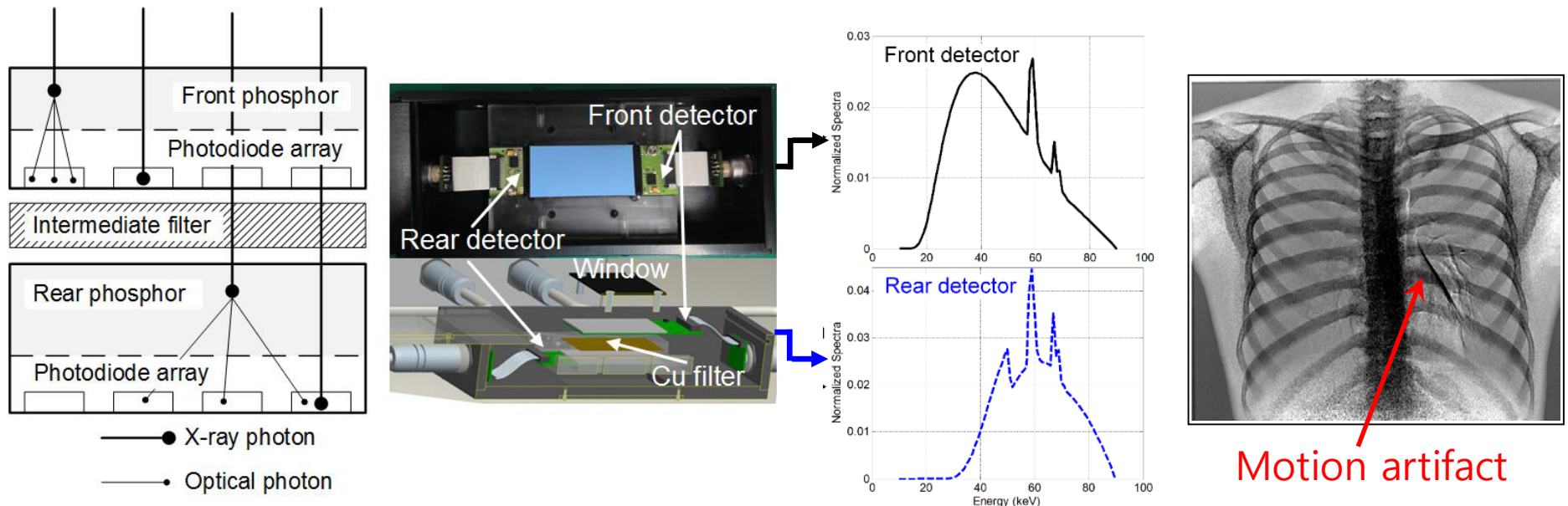
E. Shefer et al., *Curr. Radiol. Rep.* (2013)



A. Altman, R. Carmi, Philips Healthcare

Single-shot dual-energy imaging

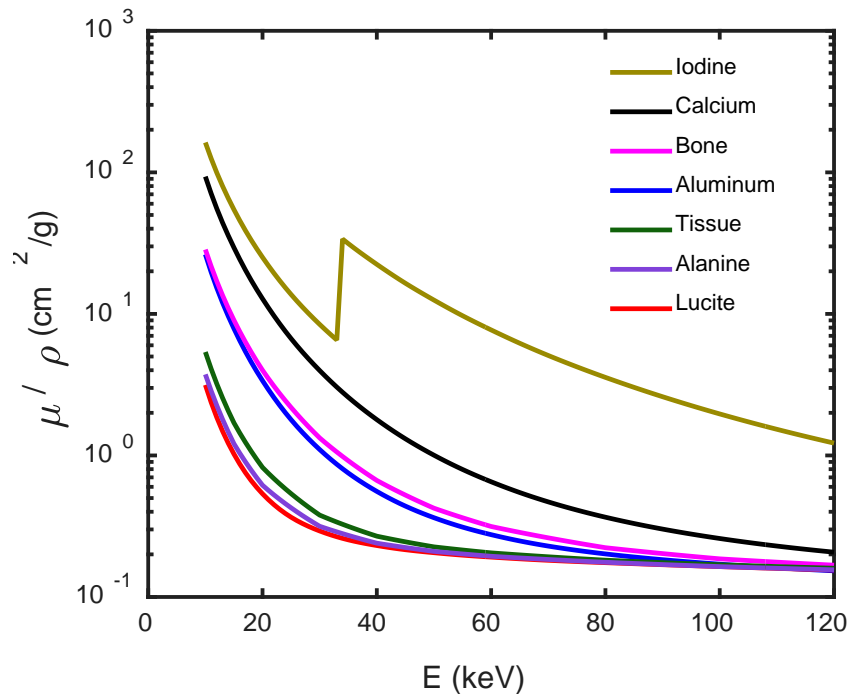
- Single-shot dual-energy (SE) imaging acquires **two images** with a **single exposure**
- With SE, motion artifacts due to heartbeat or patient motion can be avoided and irradiation dose can be reduced



Motion artifact

Attenuation coefficients

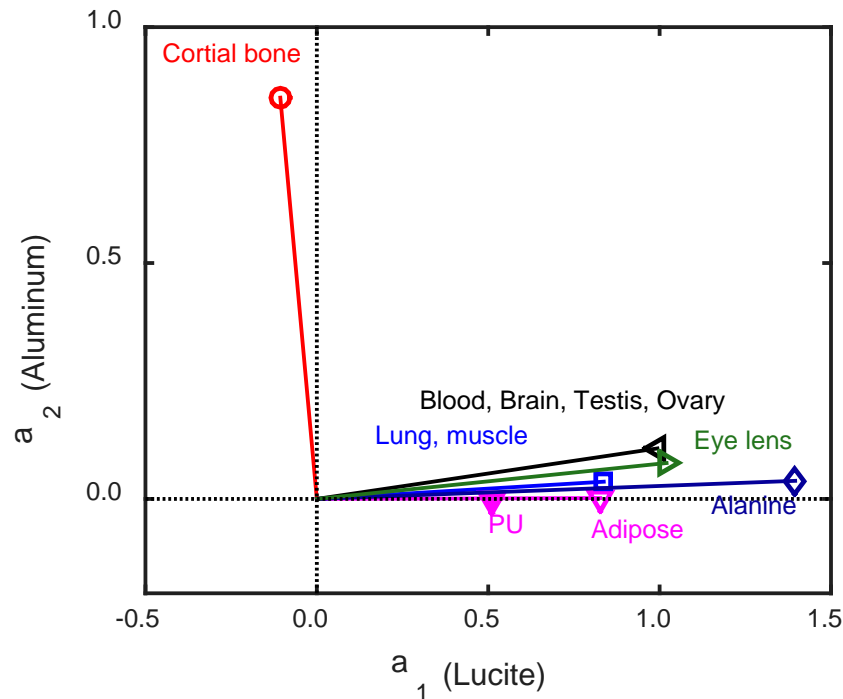
- The mass attenuation coefficients (MACs) of various materials have different values
- Particularly, materials constituting the human body and contrast agents have different MACs



| Energy [keV] | Mass attenuation coefficient [cm ² /g] | | | |
|-----------------|--|--------|-------|-------|
| | Lucite | Tissue | Al | Bone |
| 30 | 0.293 | 0.379 | 1.101 | 1.330 |
| 40 | 0.231 | 0.269 | 0.557 | 0.666 |
| 60 | 0.191 | 0.205 | 0.276 | 0.315 |
| 80 | 0.175 | 0.182 | 0.202 | 0.223 |
| 100 | 0.164 | 0.169 | 0.171 | 0.186 |
| 120 | 0.155 | 0.160 | 0.153 | 0.168 |

Basis function

- The MAC $\mu_{\xi}(E)$ of any arbitrary material ξ can be represented as a linear combination of two MAC $\mu(E)$ of basis materials
- Various materials can be coordinated using two basis materials



$$\frac{\mu_{\xi}(E)}{\rho_{\xi}} = a_1 \frac{\mu_1(E)}{\rho_1} + a_2 \frac{\mu_2(E)}{\rho_2}$$

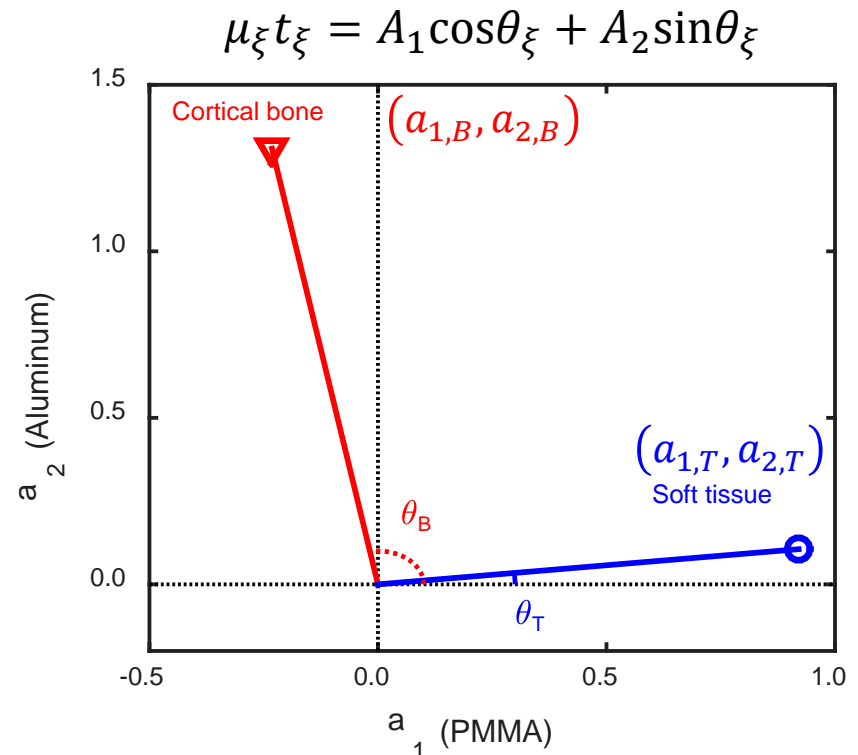
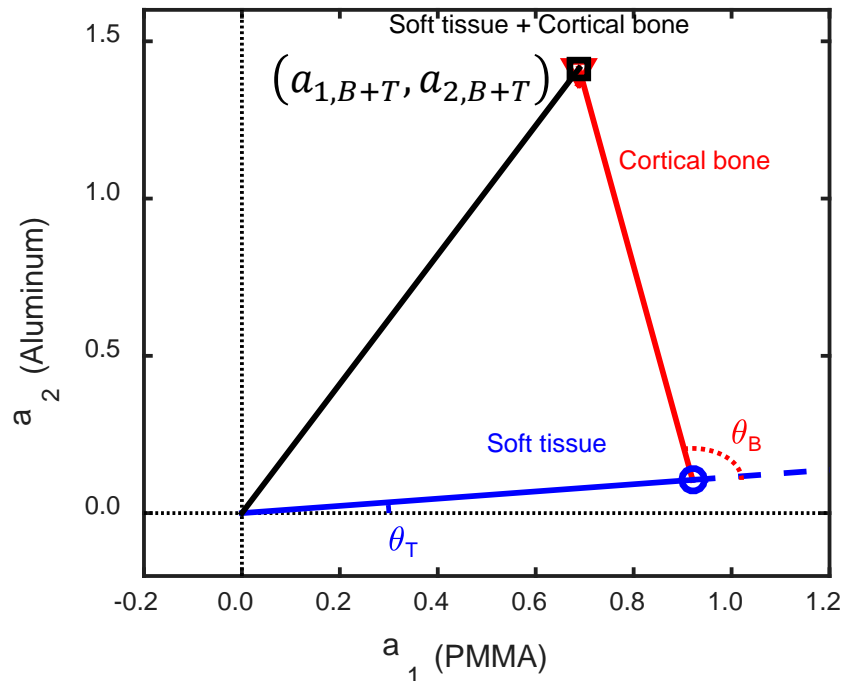
$$\begin{bmatrix} \frac{\mu_{\xi}(E_L)}{\rho_{\xi}} \\ \frac{\mu_{\xi}(E_H)}{\rho_{\xi}} \end{bmatrix} = \begin{bmatrix} \frac{\mu_1(E_L)}{\rho_1} & \frac{\mu_2(E_L)}{\rho_2} \\ \frac{\mu_1(E_H)}{\rho_1} & \frac{\mu_2(E_H)}{\rho_2} \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$$

$$\theta = \tan^{-1} \left(\frac{a_2}{a_1} \right)$$

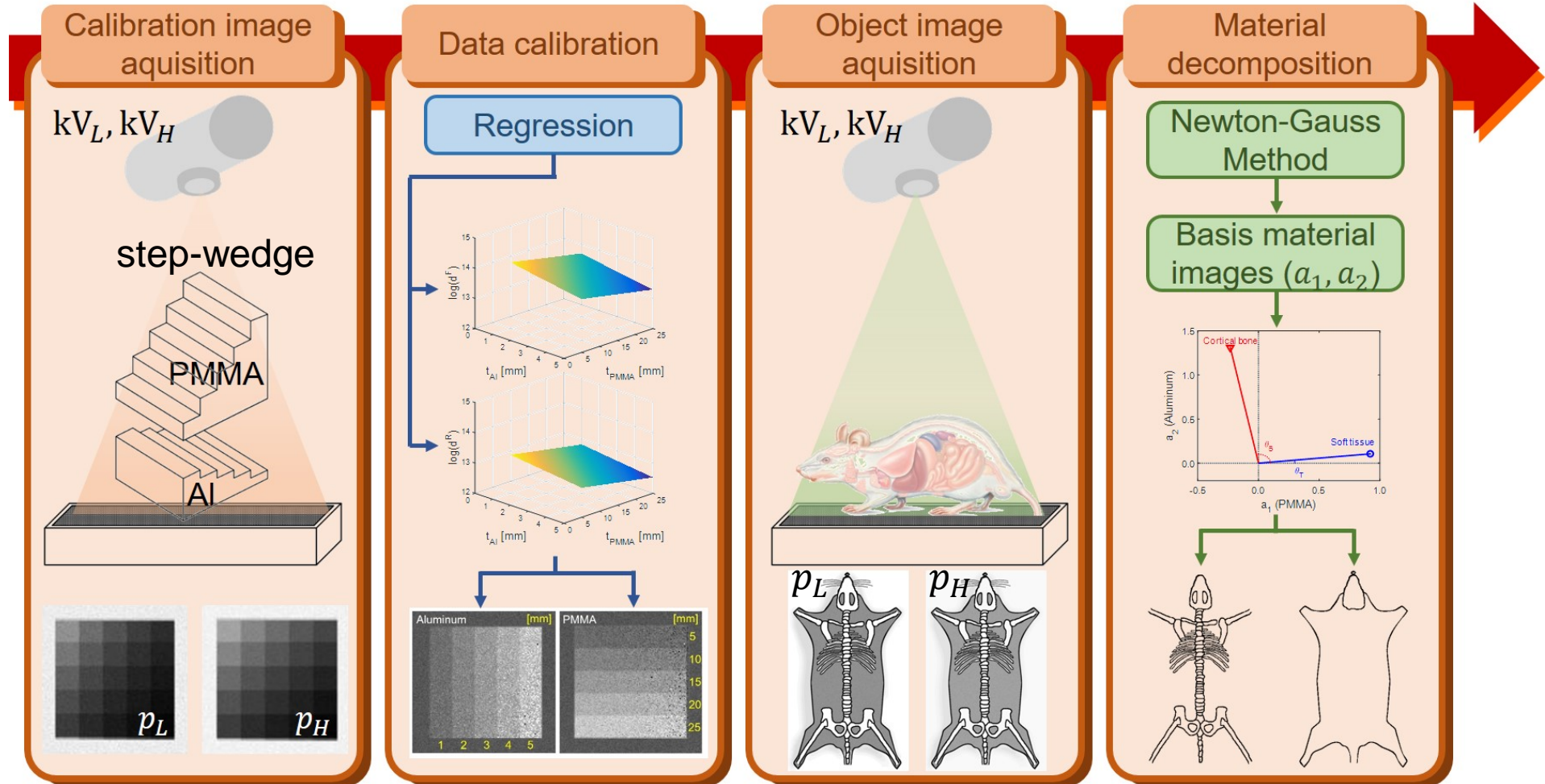
R. E. Alvarez and A. Macovski, *Phys. Med. Biol.* (1976)
L. A. Lehmann et al., *Med. Phys.* (1981)

Basis function

- The basis function vector in which the two materials are superimposed can be expressed by **sum of the basis function vectors** of each material
- The material decomposition is to identify the basis function vector describing each material



Basis material decomposition



$$p_i(A_1, A_2) = -\ln \int \phi(E_i) e^{-A_1 \mu_1(E_i) - A_2 \mu_2(E_i)} dE$$

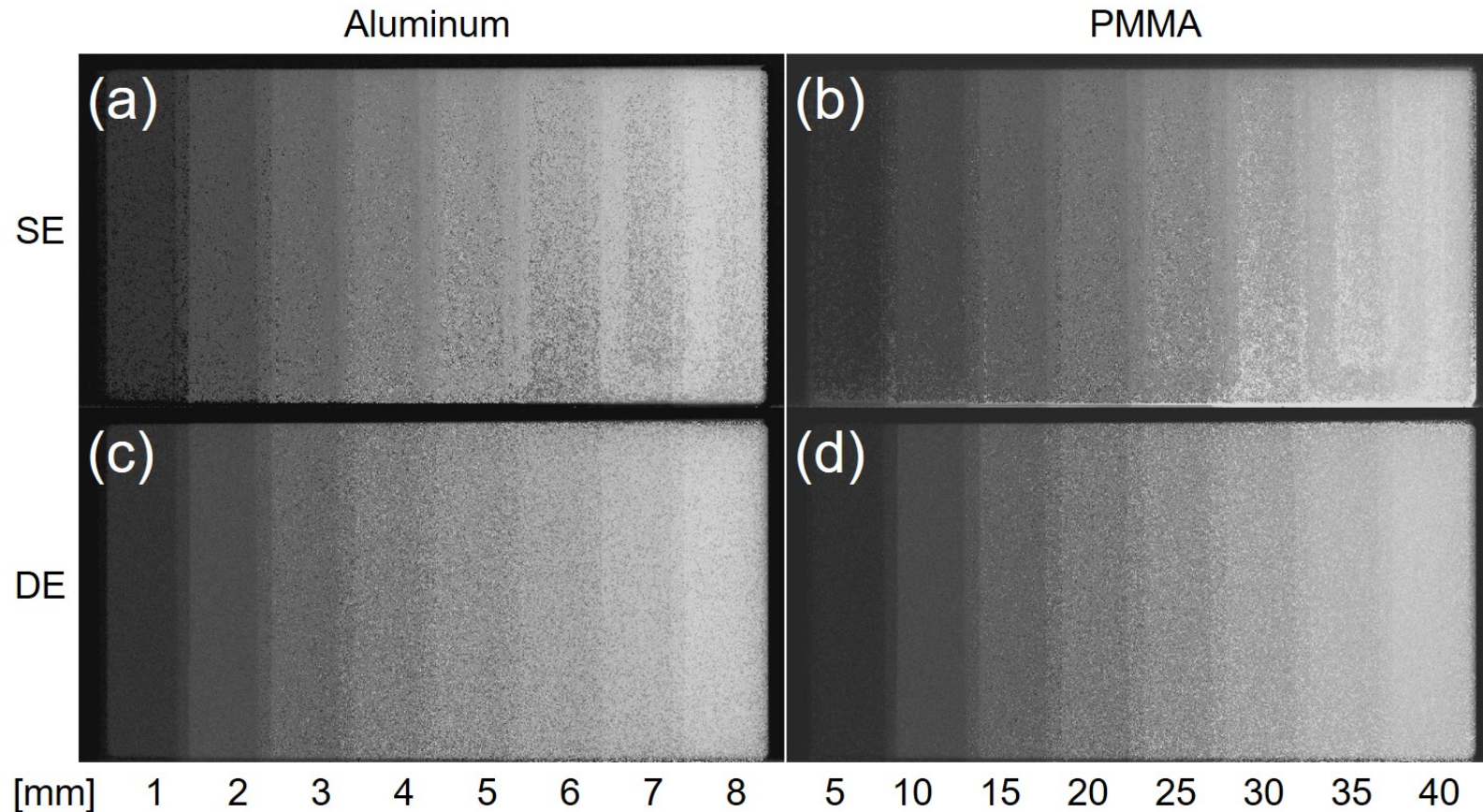
$$p_i(A_1, A_2) = b_{i1}A_1 + b_{i2}A_2 + b_{i3}A_1A_2 + b_{i4}A_1^2 + b_{i5}A_2^2$$

$$A_{j,n+1} = A_{j,n} - \left[p_L(A_{1,n}, A_{2,n}) \frac{\partial P_H}{\partial A_j} - p_H(A_{1,n}, A_{2,n}) \frac{\partial P_L}{\partial A_j} \right] / J$$

$$J = \frac{\partial P_L}{\partial A_1} \frac{\partial P_H}{\partial A_2} - \frac{\partial P_L}{\partial A_2} \frac{\partial P_H}{\partial A_1}$$

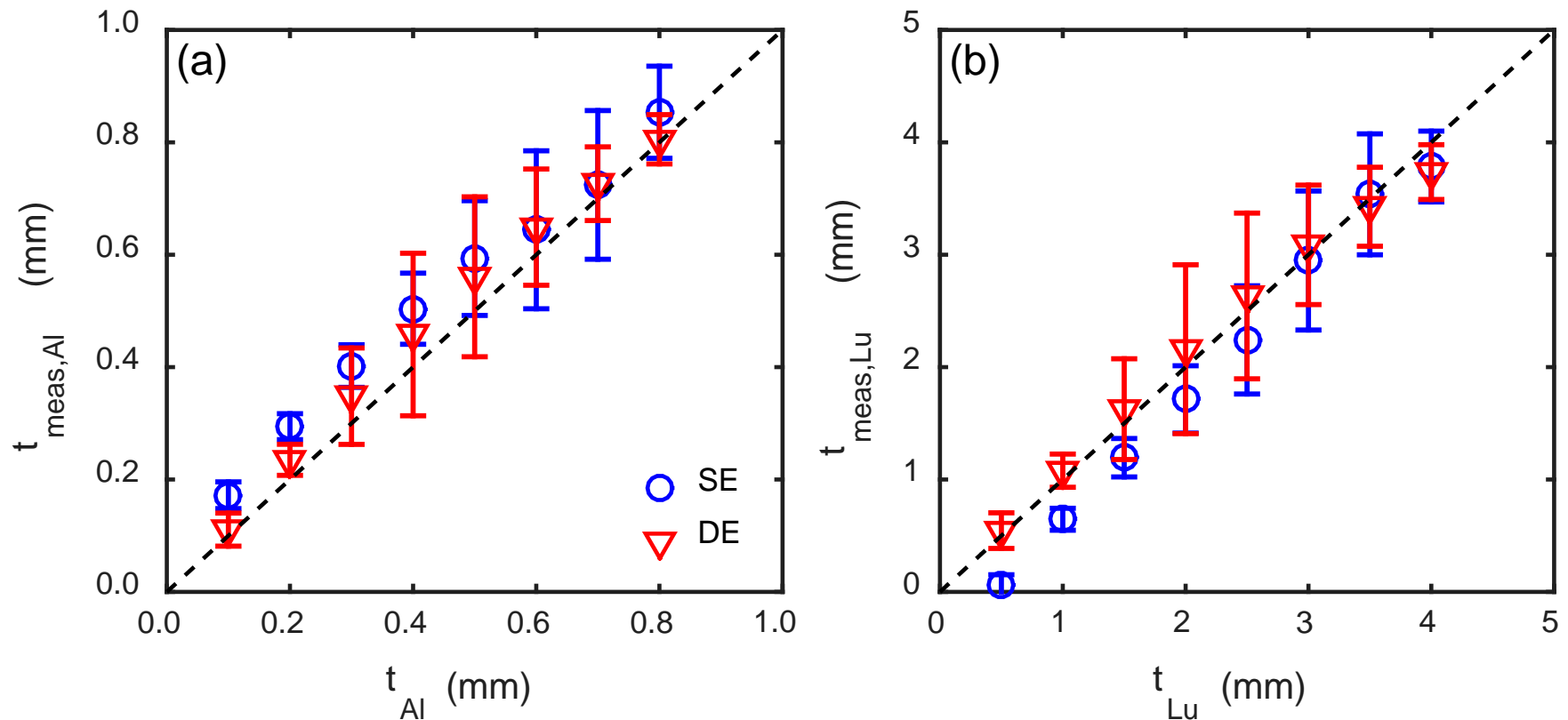
Results

- The thickness images of each material obtained using the regression analysis on the superimposed aluminum and PMMA phantom image



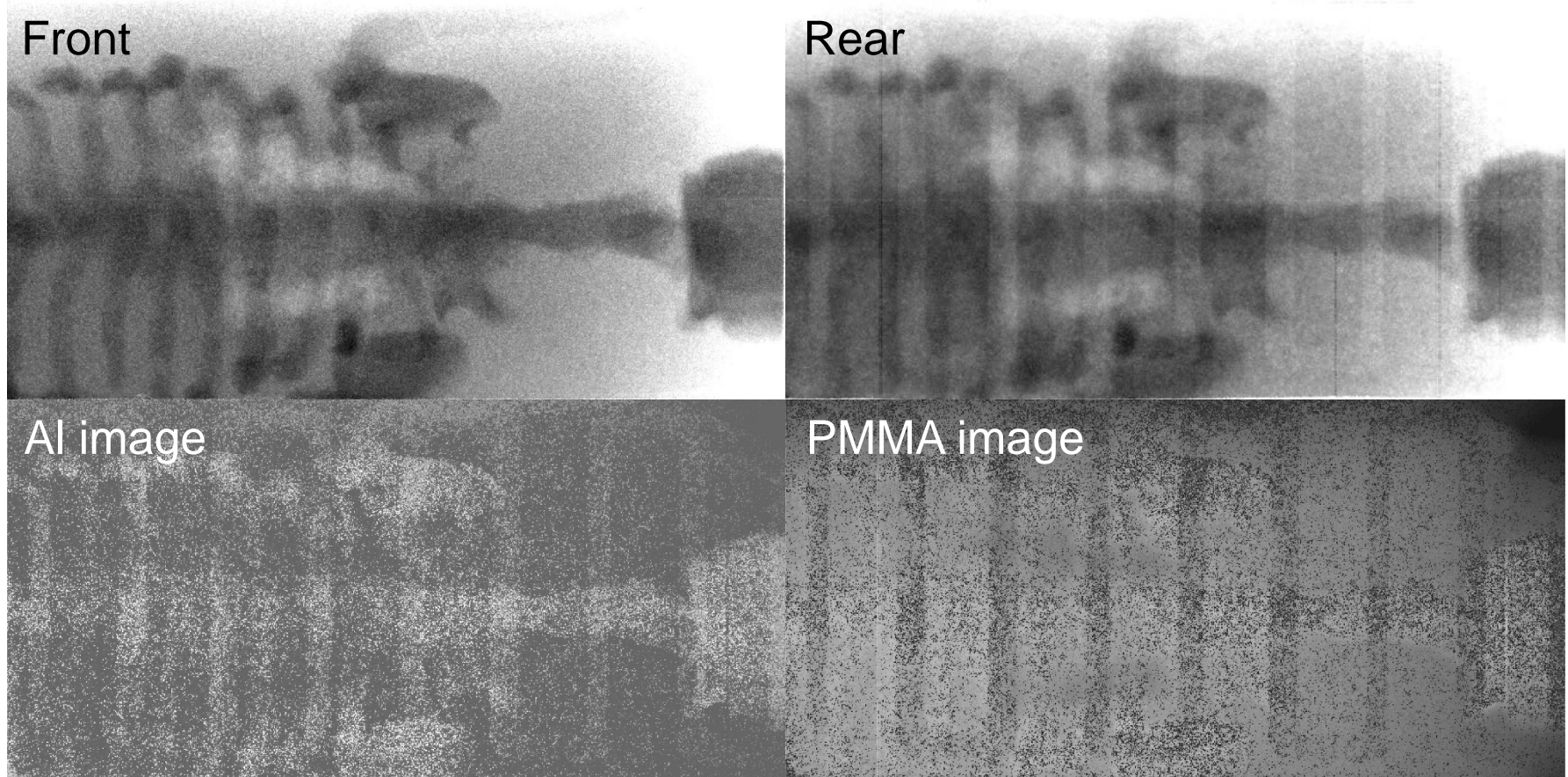
Results

- Comparison of the measured Al / PMMA thicknesses and the phantom thicknesses
- The phantom thickness is reasonably estimated from the decomposed images. However, DE is better than SE



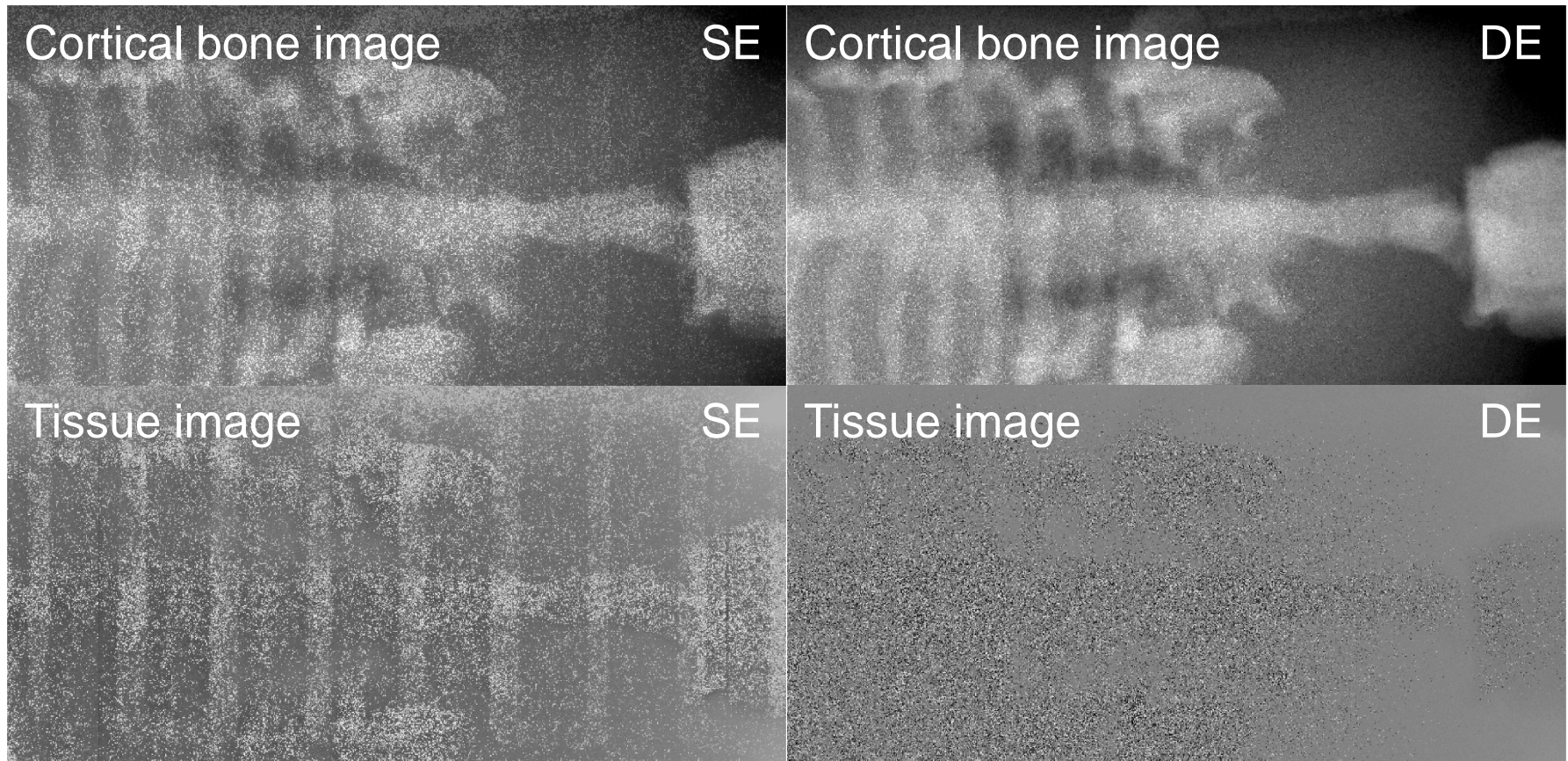
Results

- Basis material images obtained for the mouse phantom image using the sandwich detector



Results

- Cortical bone and tissue images using the basis function
- DE shows a better performance than SE, but SE is **feasible for the material decomposition imaging**



Conclusion

- **Basis material decomposition** improves the **conspicuity** by removing the background clutters by vector operation using the **basis function**
- The projection signal of the **wedge phantom** can be described by the polynomials of two material thicknesses (from **the calibration**)
- Pixel signal of an **arbitrary object can be expressed as the linear combination** of two material thicknesses by solving the inverse of the polynomials
- We have validated the **linearity** of material-specific pixel signal using the wedge phantom
- We have verified the applicability of **material decomposition radiography** using the sandwich detector to **bone and soft-tissue images** of a mouse-mimetic phantom
- **SE results** show a **sufficient feasibility**