

KNS Fall Meeting

Linear Analysis of X-ray Imaging

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Ho Kyung Kim

hokyung@pusan.ac.kr

Pusan National University

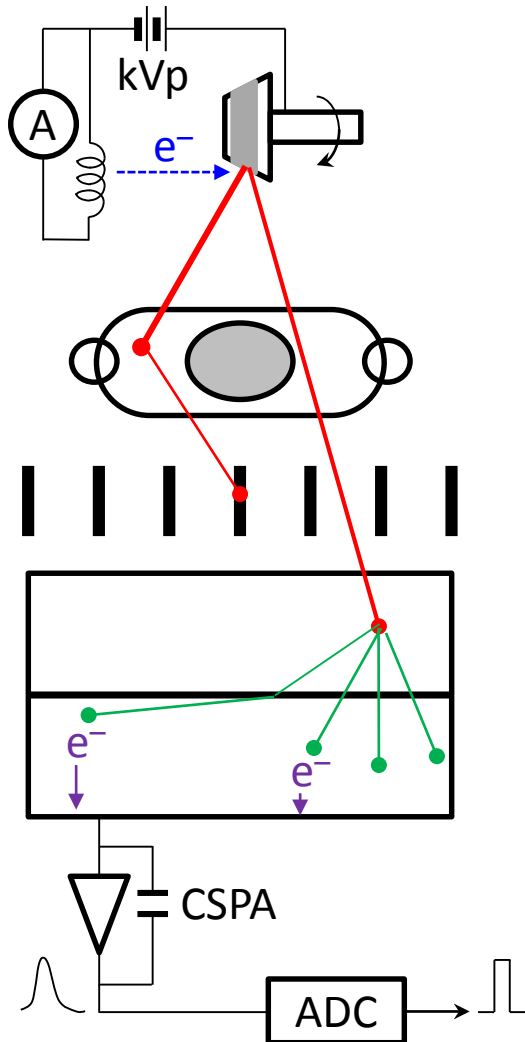
Linear analysis on:

- Image quality
 - Fourier-based image quality

- Detector performance
 - Cascaded linear-systems theory on flat-panel detectors

- System performance
 - Dual-energy CBCT

Imaging chain



- X-ray tube
 - Spectrum (polyenergetic)
 - Focal spot size (image blur)
 - Heel effect (nonuniformity)
- Patient
 - Scatter (the most harmful)
- Anti-scatter grid
 - Transmittance (or selectivity)
- Detector
 - Converter
 - Glare
 - Photodiode
 - Amplifier
 - ADC

Image quality

- Contrast

- Signal difference = $\Delta d = d_b - d_s$
- Contrast = $\frac{\Delta d}{d_b}$

- Noise

- *Standard deviation* of signal
- $$\text{CNR} = \frac{\Delta d}{\sqrt{\frac{\sigma_b^2 + \sigma_s^2}{2}}}$$

- Spatial resolution

- FWHM

- Artifact

- Scatter

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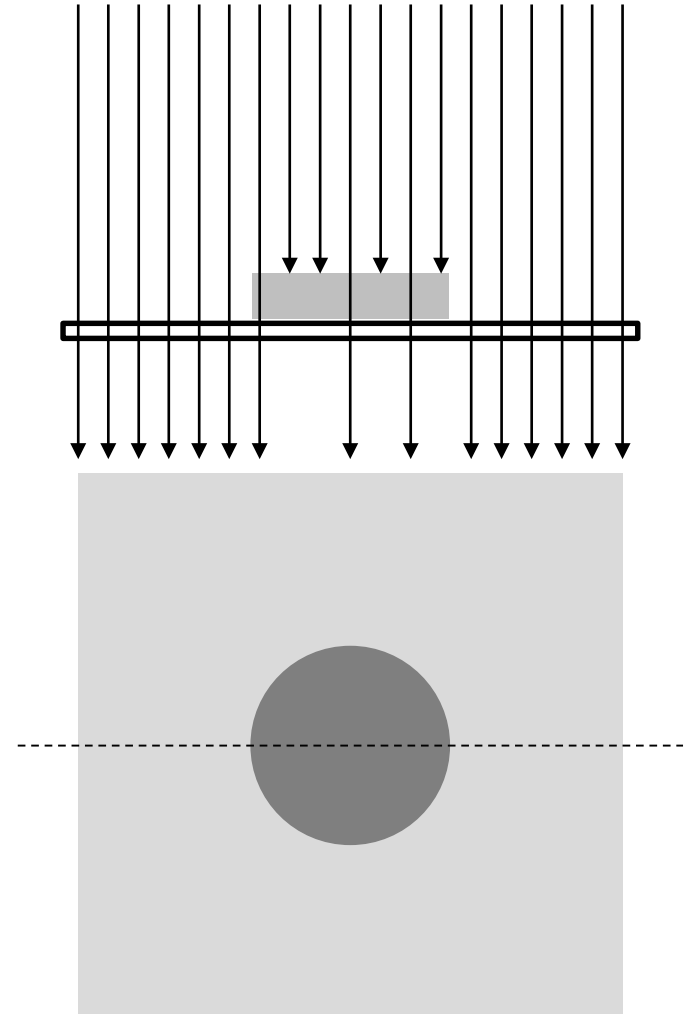
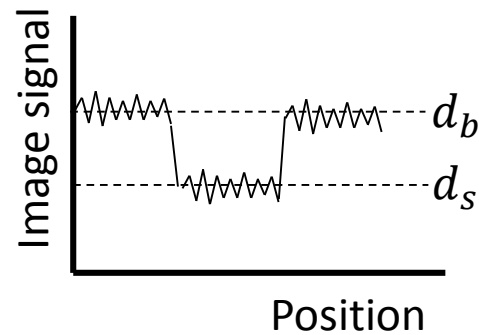


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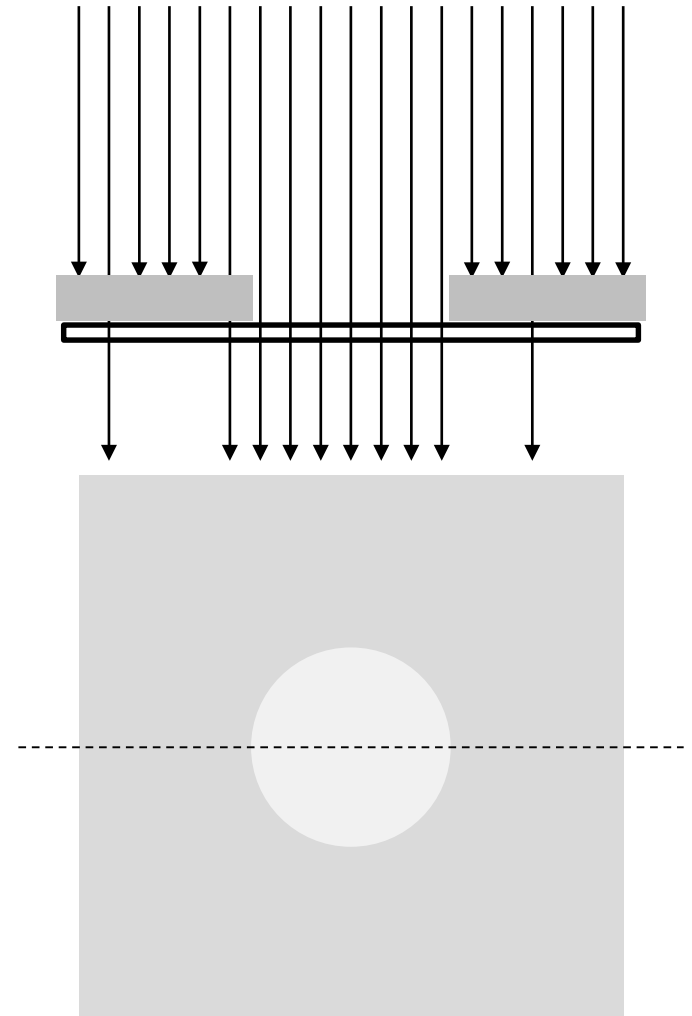
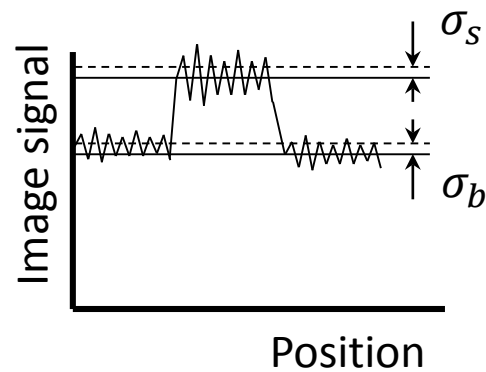


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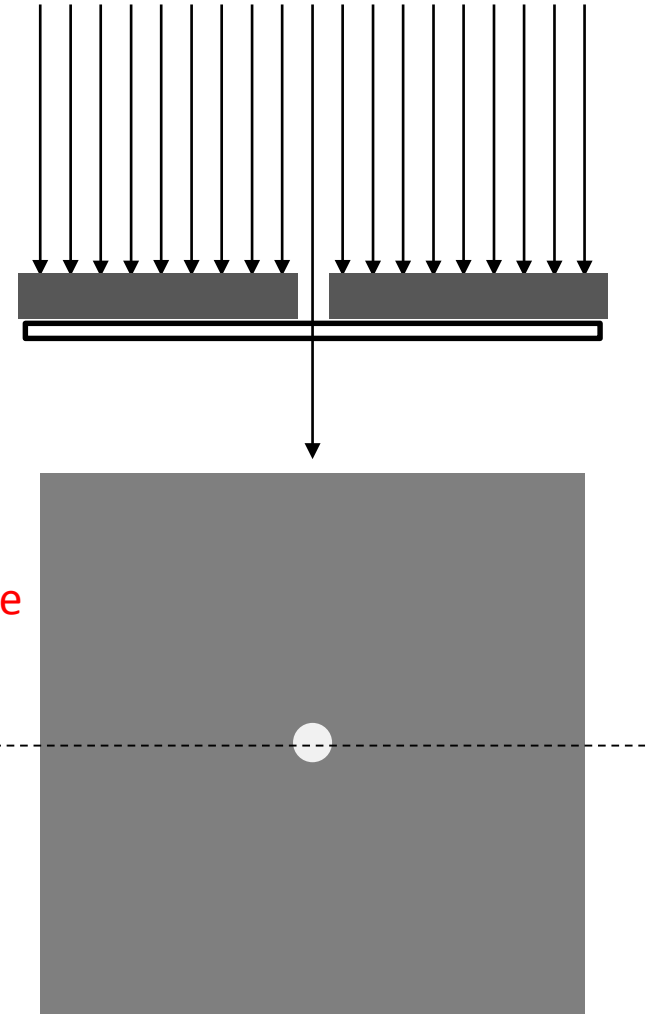
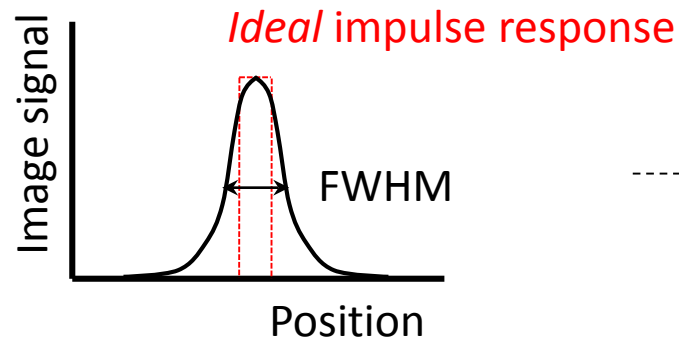


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Spatial correlation



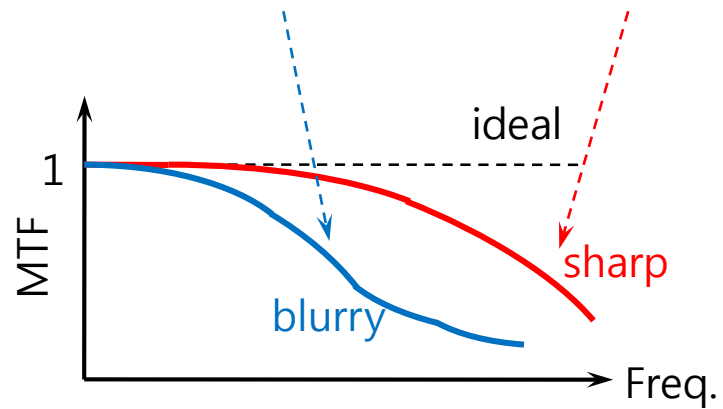
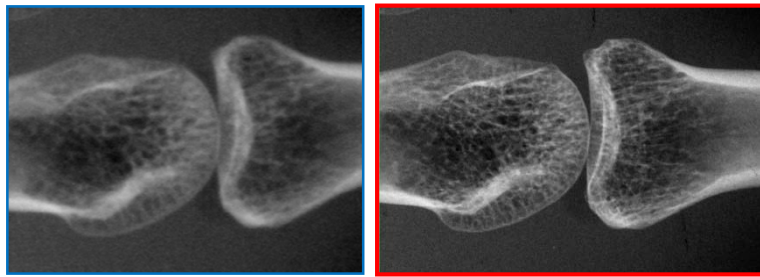
- These two images have the same *pixel variance*, but different *correlation structure* (different textures!)
- Simple image pixel variance ignores second-moment statistics (correlation between pixels)

(Images taken from) R. F. Wagner | AAPM | 2004

Fourier-based metrics

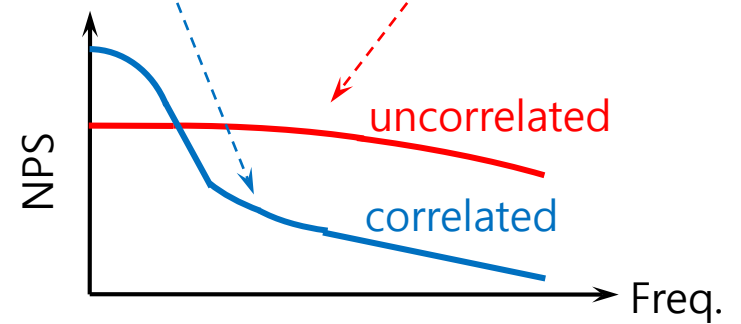
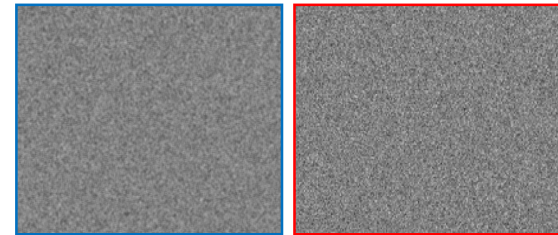
- Particle-based metrics

- contrast transfer
- noise variance



- Fourier-based metrics

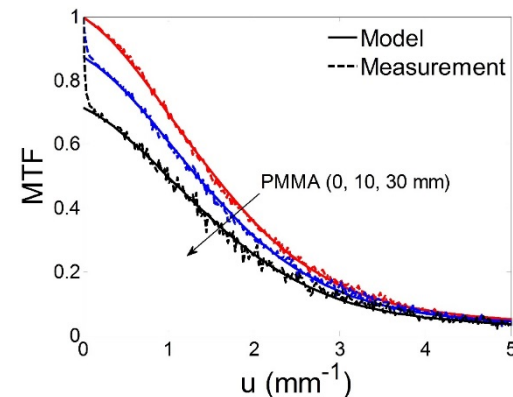
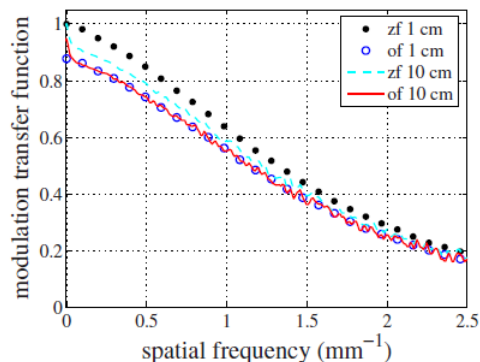
- modulation-transfer function
- Wiener noise-power spectrum



MTF

$$\text{MTF}(\mathbf{u}) = \mathcal{F} \left\{ \frac{\text{psf}(\mathbf{x})}{\int_{-\infty}^{\infty} \text{psf}(\mathbf{x}) d\mathbf{x}} \right\} = \frac{|T(\mathbf{u})|}{|T(\mathbf{0})|}$$

- The conventional zero-frequency normalization may result in inflated MTF values ¹⁾
 - Consider the analyzing ROI size enough to take the optical glare into
- Scatter x-ray photons also reduce the zero-frequency MTF values ²⁾



¹⁾ S. N. Friedman and I. A. Cunningham | Med. Phys. | 2008

²⁾ J. Park et al. | SPIE | 2016

NPS

$$\text{NPS}(\mathbf{u}) = \frac{1}{\Delta \mathbf{u}} \langle |\mathcal{F}\{\Delta d(\mathbf{x})\}|^2 \rangle$$

- How to determine $\text{NPS}(\mathbf{0})$; hence $\text{DQE}(\mathbf{0})$?

$$\text{NPS}(\mathbf{0}) = A_{eff} \sigma_d^2$$

$$\text{DQE}(\mathbf{0}) = \frac{\bar{d}^2}{\bar{q}_0 A_{eff} \sigma_d^2}$$

Requirements

1. Linearity

- image intensity scales with x-ray input

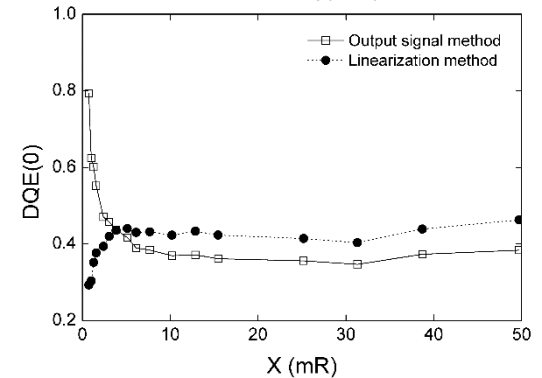
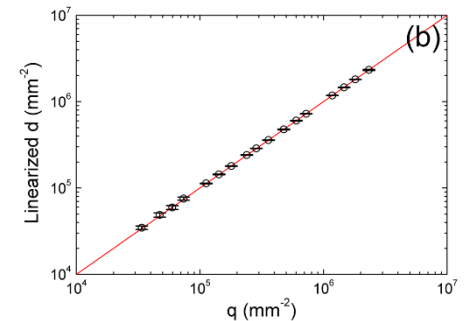
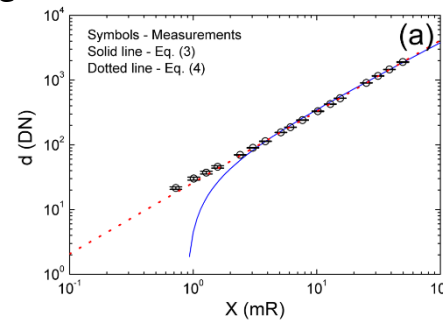
2. Shift invariance

- impulse-response function is same over image

3. Stationary noise

■ If either fails:

- linearization
- small-signal approach
- regional analysis



J. C. Han *et al.* | JKPS | 2014

Forward model

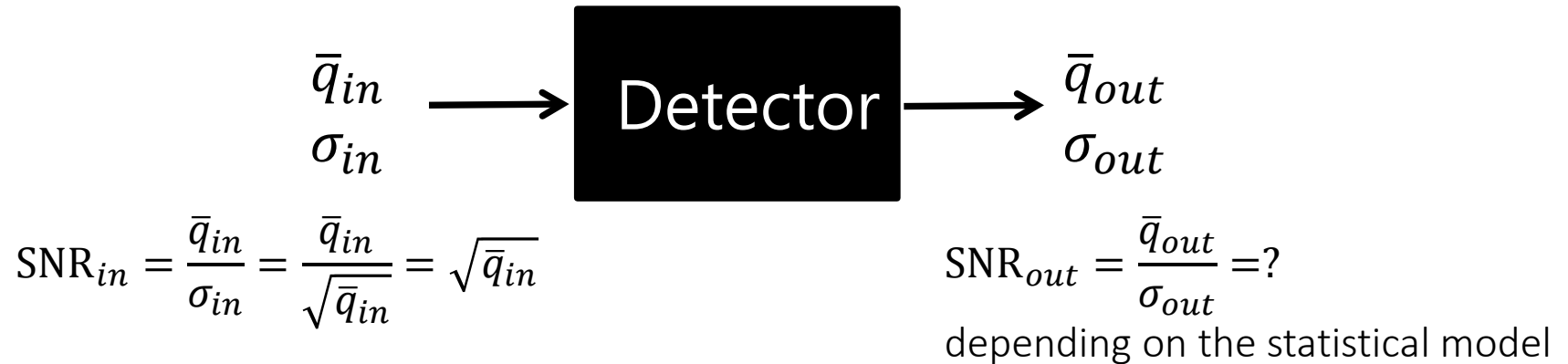
$$d(\xi; s) = Xka^2(1 + \text{SPR}) \int_0^\infty q_0(E) e^{-\int_s \mu(\mathbf{x}, E) ds} R(E) dE$$

where

$$R(E) = \int_0^L e^{-\mu_d(E)z} \mu_E(E) g(E) dz = (1 - e^{-\mu_d(E)L}) \frac{\mu_E(E)}{\mu_d(E)} g(E)$$

- $g(E)$
 - linear approximation by using the cascaded linear-systems theory

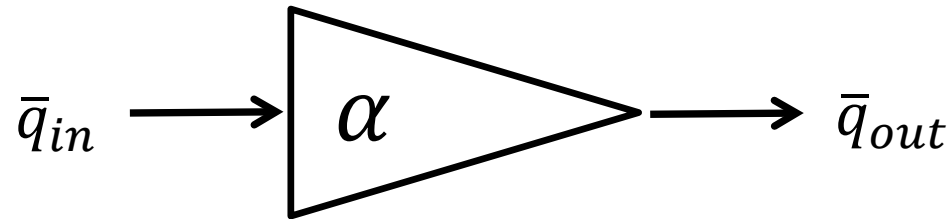
Signal & noise transfer



- Detective quantum efficiency
 - working even for different units between *in* and *out*

$$DQE = \frac{SNR_{out}^2}{SNR_{in}^2} \text{ (called } \textit{conceptual} \text{ or } SNR^2\text{-transfer form)}$$

Binomial quantum-detection detector (*detection or not*)



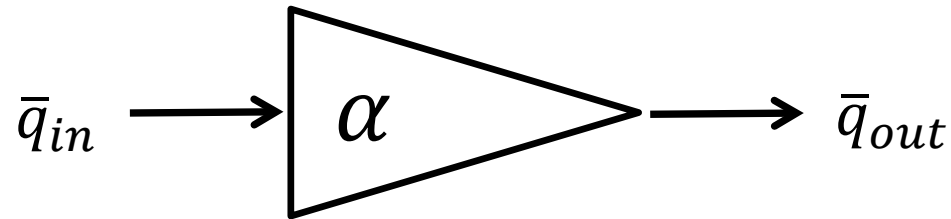
$$\bar{q}_{out} = \alpha \bar{q}_{in}$$

$$\sigma_{out}^2 = \alpha^2 \sigma_{in}^2 = \alpha^2 \bar{q}_{in}$$

$$\text{SNR}_{out} = \frac{\bar{q}_{out}}{\sigma_{out}} = \frac{\alpha \bar{q}_{in}}{\alpha \sqrt{\bar{q}_{in}}} = \sqrt{\bar{q}_{in}}$$

$$\text{DQE} = \frac{\text{SNR}_{out}^2}{\text{SNR}_{in}^2} = 1$$

Binomial quantum-detection detector (*detection or not*)



$$\bar{q}_{out} = \alpha \bar{q}_{in}$$

$$\sigma_{out}^2 = \alpha^2 \sigma_{in}^2 + \bar{q}_{in} \sigma_g^2 \text{ } ^{1)} = \alpha^2 \bar{q}_{in} + \bar{q}_{in} \alpha (1 - \alpha) = \alpha \bar{q}_{in}$$

$$\text{SNR}_{out} = \frac{\bar{q}_{out}}{\sigma_{out}} = \frac{\alpha \bar{q}_{in}}{\sqrt{\alpha \bar{q}_{in}}} = \sqrt{\alpha \bar{q}_{in}}$$

$$\text{DQE} = \frac{\text{SNR}_{out}^2}{\text{SNR}_{in}^2} = \alpha$$

¹⁾ M. Rabbani, R. Sahw, and R. L. Van Metter | JOSA | 1987

Various forms of DQE

■ Descriptive

- In terms of parameters determined from measured images

$$\text{DQE}(u) = \frac{\bar{q}^2 |\text{GMTF}(u)|^2 / \text{NPS}(u)}{\bar{q}} = \frac{\bar{q} G^2 \text{MTF}(u)^2}{\text{NPS}(u)} = \frac{\text{MTF}(u)^2}{\bar{q} [\text{NPS}(u)/d^2]}$$

■ Stochastic (most general)

- NPS by a *deterministic* syst. relative to an actual *stochastic* syst.

$$\text{DQE}(u) = \frac{\text{NPS}_{\text{ideal}}(u)}{\text{NPS}_{\text{actual}}(u)} = \frac{\text{NPS}_{\text{in}}(u) |\text{GMTF}(u)|^2}{\text{NPS}(u)} = \frac{\bar{q} G^2 \text{MTF}(u)^2}{\text{NPS}(u)}$$

■ Predictive

- In terms of known design parameters

$$\text{DQE}(u) = \frac{1}{1 + \sum_{j=1}^M \frac{1 + \varepsilon g_j \text{MTF}_j^2(u)}{\prod_{i=1}^j \bar{g}_i \text{MTF}_i^2(u)}}$$

Flat-panel detector

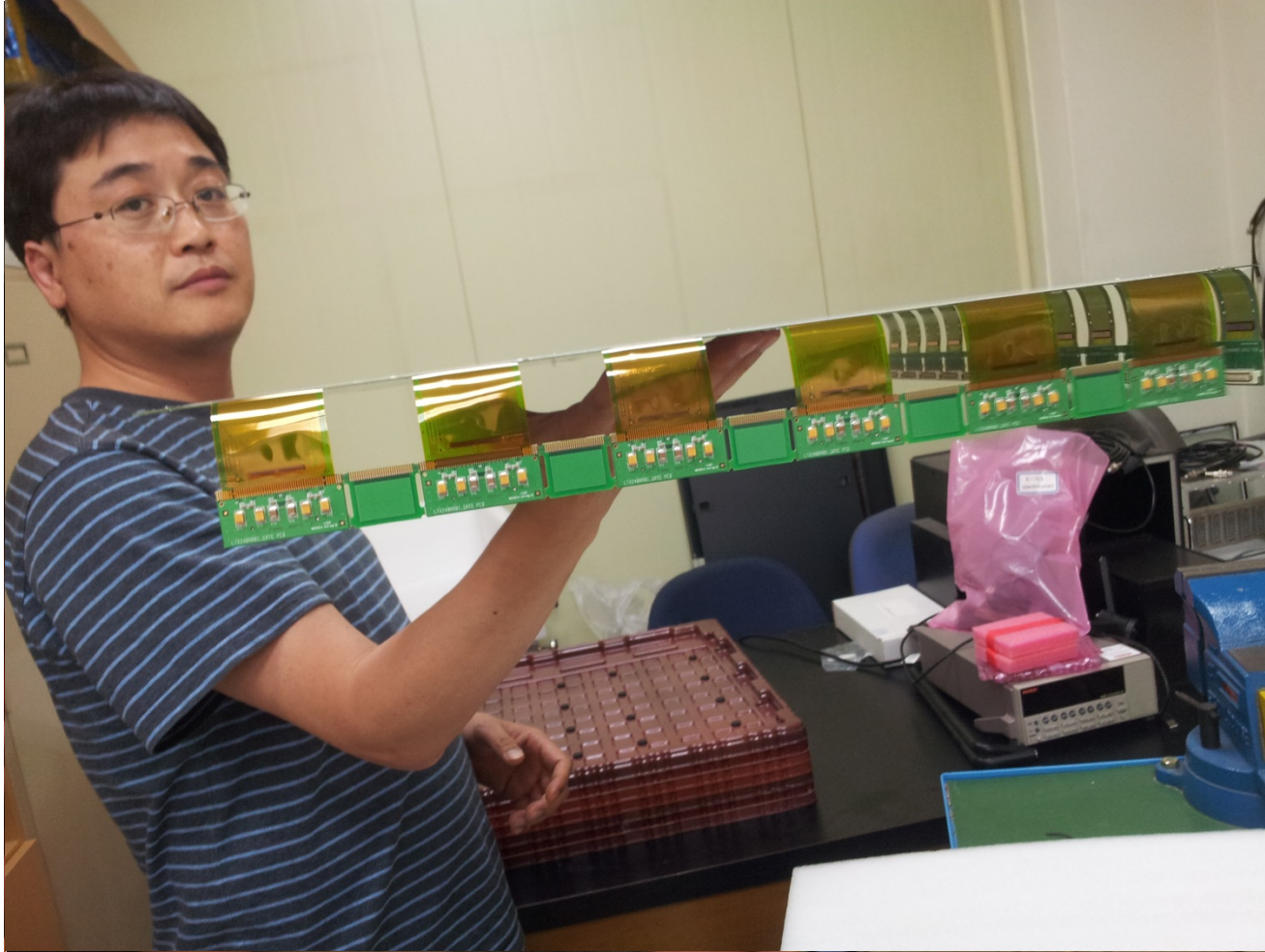
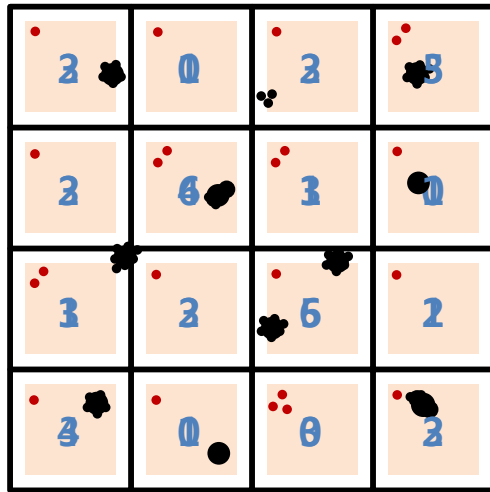
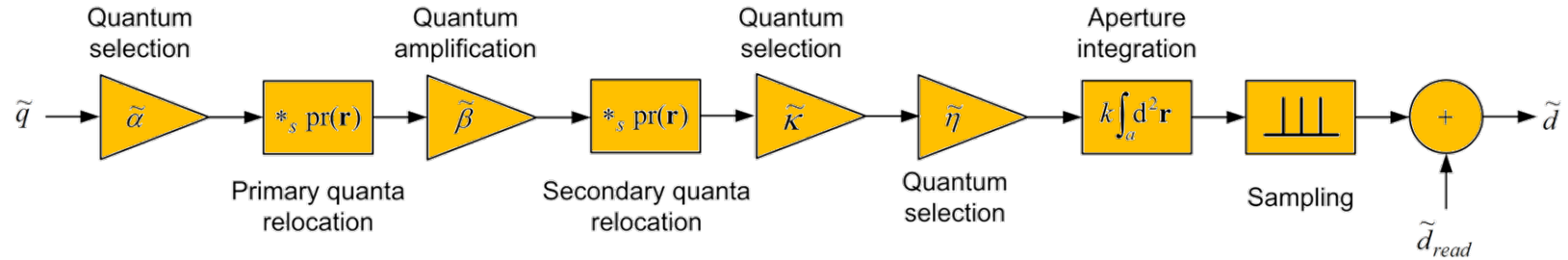
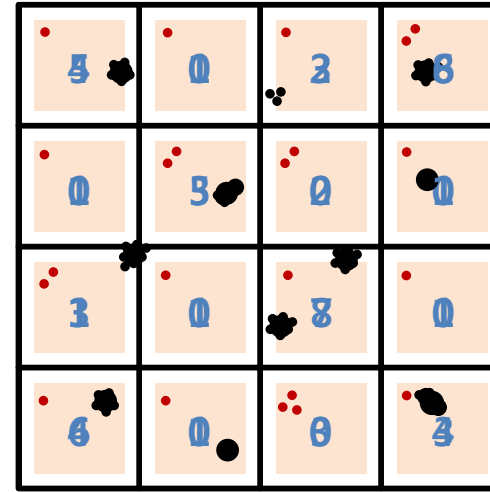


Image Courtesy of Samsung Electronics, Co., Ltd. & PNU

Cascaded linear-systems model



Long-range scattering



Short-range scattering

Predictive DQE based on CSA

$$\text{DQE}(\mathbf{k}) = \frac{T^2(\mathbf{k})\text{sinc}^2(a\mathbf{k})}{\frac{1}{\alpha\beta\kappa\eta} \left[\frac{1}{\gamma} + \kappa\eta\left(\frac{\beta}{I}-1\right) \sum_{j=0}^{\infty} \left\{ T^2\left(\mathbf{k} \pm \frac{j}{p}\right)\text{sinc}^2\left(a\left(\mathbf{k} \pm \frac{j}{p}\right)\right) \right\} \right]} + \frac{\sigma_{add}^2}{\gamma\bar{q}a^2(\alpha\beta\kappa\eta)^2}$$

- Dose-independent if only if the additive noise can be ignored
- Additive noise is harmful to DQE at high frequencies where the number of secondary quanta lessens

Implications

$$\frac{\sigma_{add}^2}{\gamma \bar{q} a^2 (\alpha \beta \kappa \eta)^2} \rightarrow 0$$

- $\sigma_{add} \downarrow$
 - new metal line process
- $\gamma \uparrow$
 - limited by the TFT design rule
 - critical to high-resolution FPD (e.g. α -Se)
 - Electrostatic lens design
- $\bar{q} a^2 \uparrow$
 - wrong approach (\because patient dose \uparrow)
- $\alpha \uparrow$
 - high Z converters
 - thick converters \Rightarrow $MTF(u) \downarrow$
- $\beta \uparrow$
 - converters having a lower W-value
 - e.g. CdZnTe, $HgI_2 > 10 \times \alpha$ -Se
- $\kappa \eta \uparrow$
 - block small leakages (optical and charge leakages)
 - optical mismatch, poor charge-collection efficiency ...

Forward model, again

$$d(\xi; s) = Xka^2 \int_0^\infty q_0(E) e^{-\int_s \mu(\mathbf{x}, E) ds} R(E) dE$$

- $\mu(\mathbf{x}, E)$ is implicitly carved in projection signal
 - *spatial averaging*
 - hiding lesions
 - resulting in the background noise clutter ($\sigma_{anat} \geq 10 \times \sigma_q$)
 - *energy averaging*
 - same $\hat{\mu}$ from different materials (ρ, Z)

- Consequently, projection radiography provides poor lesion conspicuity

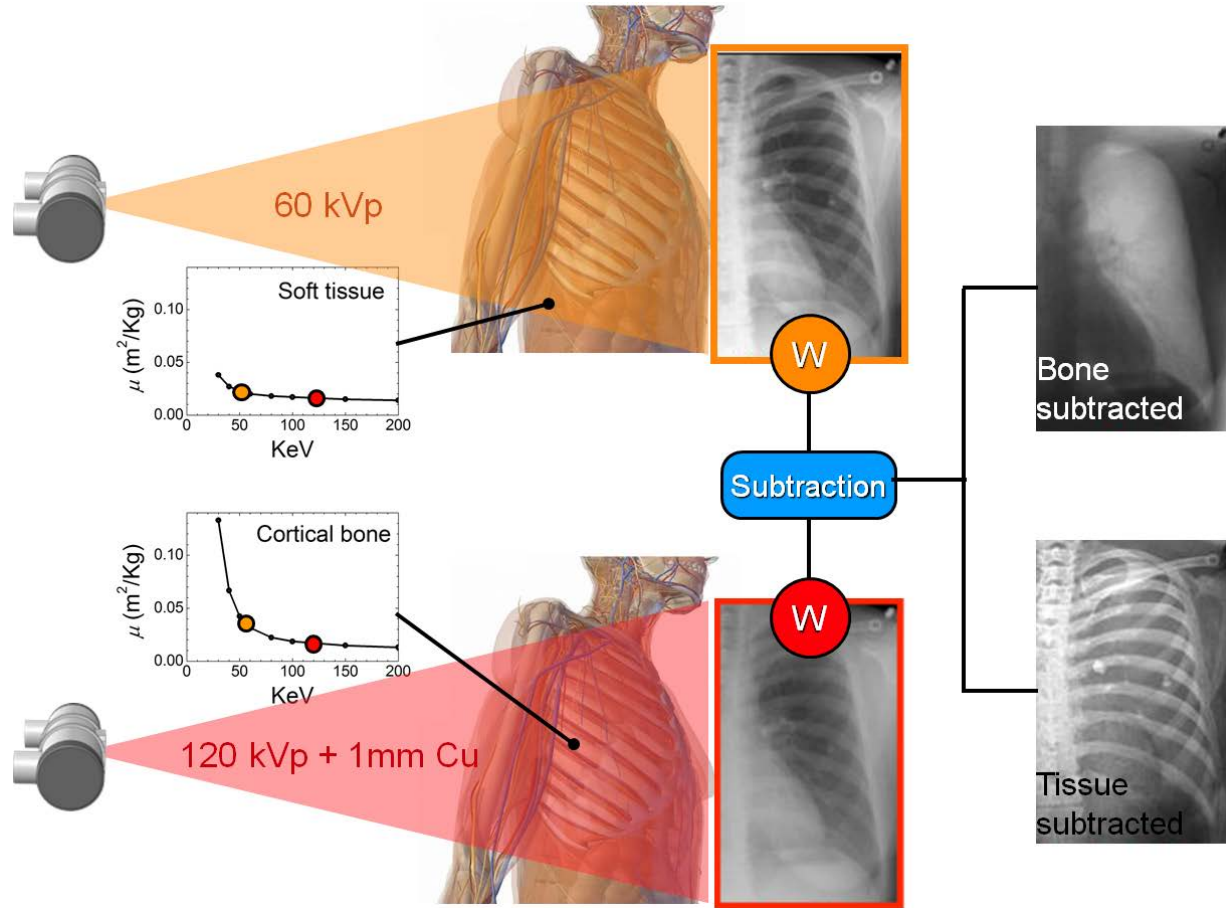
Space discrimination

Monochromatic approximation

$$p(\xi) = -\log \frac{d(\xi)}{\bar{d}_0} \approx \int_s \mu(\mathbf{x}, E_{eff}) ds$$

$$\hat{f}(\mathbf{x}) = a \int_{\theta_{min}}^{\theta_{max}} \hat{p}(\xi; \theta) \Big|_{\xi=\mathbf{x}\cdot\boldsymbol{\theta}} d\theta$$

Energy discrimination



Linear approximation

$$p(\xi; s; \text{kVp}) \approx \sum_j \hat{\mu}_j(\text{kVp}) \int_s f_j(\mathbf{x}) ds$$

where

$$\hat{\mu}(\text{kVp}) = \mathbb{E} \left\{ \frac{\int_0^{\text{kVp}} q_0(E) R(E) \mu(E) dE}{\int_0^{\text{kVp}} q_0(E) R(E) dE} \right\}$$

- Two-basis (i.e., $j = 2$) material analysis (e.g., bone & soft tissue)

$$p(\text{kVp}) = \hat{\mu}_b(\text{kVp})t_b + \hat{\mu}_s(\text{kVp})t_s$$

$$t_j = \bar{p}(p_{\text{kVp}_H}) \pm w_j p(p_{\text{kVp}_L})$$

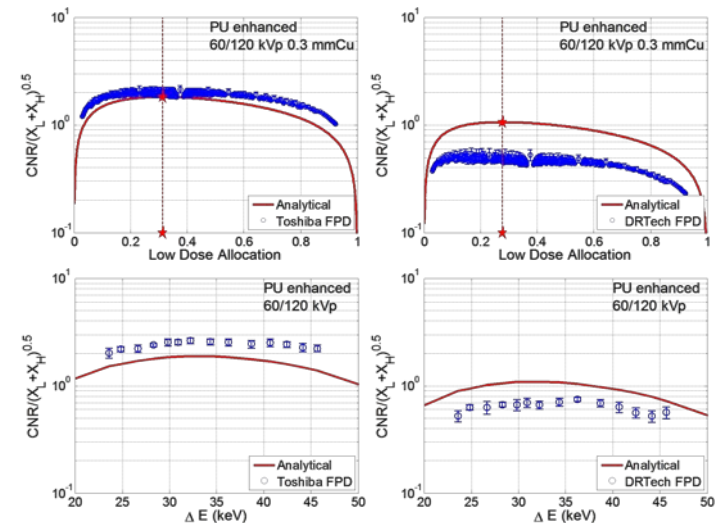
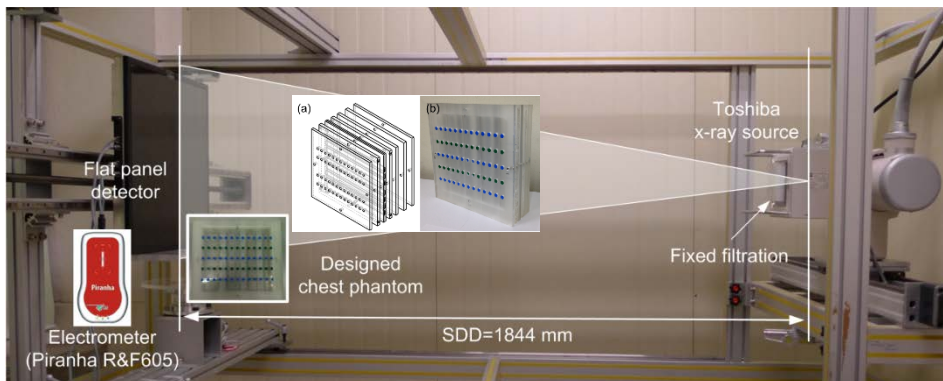
Optimization

$$\theta^* = \arg \max_{\theta} f(\theta)$$

- Ex) Determine the optimal exposure fraction in low-energy imaging which maximizing the benefit-to-cost performance

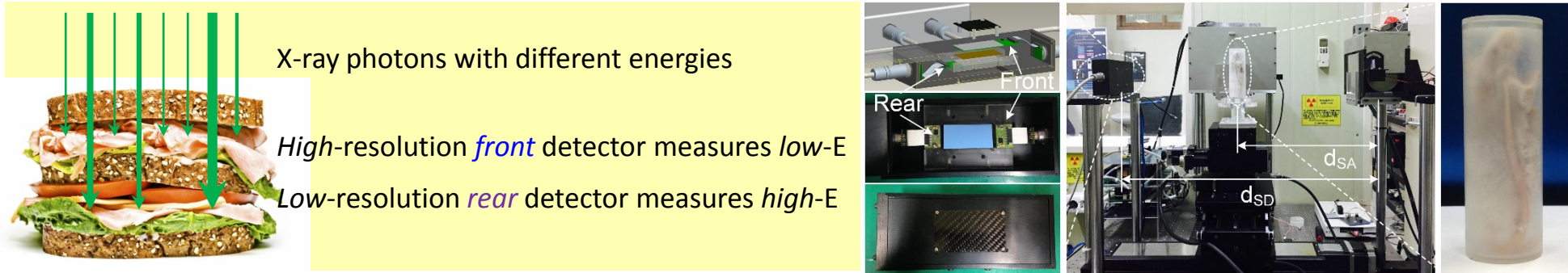
$$f(A_{XL}) = \frac{\text{CNR}_j^2}{X} = C_j^2 \left[\frac{1}{\bar{q}_{H0}(1 - A_{XL})\text{DQE}_H} + \frac{w_j^2}{\bar{q}_{L0}A_{XL}\text{DQE}_L} \right]^{-1}$$

$$A_{XL}^* = \left[1 + \frac{1}{w_j} \sqrt{\frac{\bar{q}_{L0}\text{DQE}_L}{\bar{q}_{H0}\text{DQE}_H}} \right]^{-1}$$

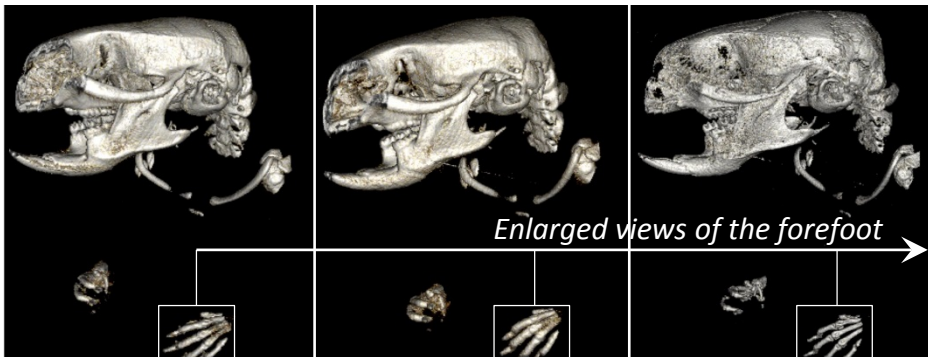


Manuscript in preparation

Single-shot DEI



Modulation transfer functions



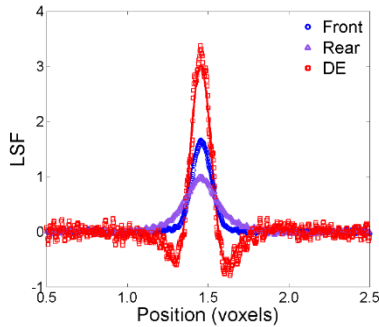
Imaging theory

$$p(\xi) = \bar{\lambda}q(\xi) * g(\xi) = \bar{\lambda}\bar{q}\bar{g}L(\xi)$$

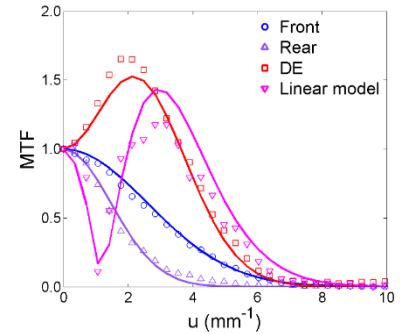
$$\hat{p}(\xi) = \bar{\lambda}\bar{q}\bar{g}L(\xi) * h(\xi) = \bar{\lambda}\bar{q}\bar{g}\hat{L}(\xi)$$

$$\hat{f}(\mathbf{x}) = a \int_0^{2\pi} \hat{p}(\xi; \theta) \Big|_{\xi=\mathbf{x}\cdot\boldsymbol{\theta}} d\theta = a\bar{\lambda}\bar{q}\bar{g} \int_0^{2\pi} \hat{L}(\mathbf{x}\cdot\boldsymbol{\theta}; \theta) d\theta$$

$$\text{MTF}(\mathbf{u}) = \mathcal{F} \left\{ \frac{\hat{f}(\mathbf{x})}{\int_{-\infty}^{\infty} \hat{f}(\mathbf{x}) d\mathbf{x}} \right\} = \mathcal{F} \left\{ \int_0^{2\pi} \hat{L}(\mathbf{x}\cdot\boldsymbol{\theta}; \theta) d\theta \right\}$$



$$\text{MTF}_{DE}(\mathbf{u}) = \frac{\bar{w}\text{MTF}_{DE}(\mathbf{u}) - \text{MTF}_{DE}(\mathbf{u})}{\bar{w} - 1}$$

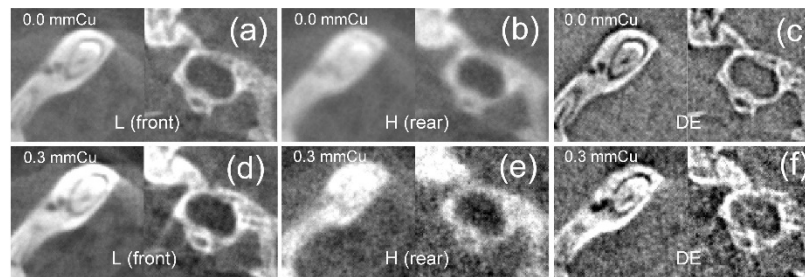
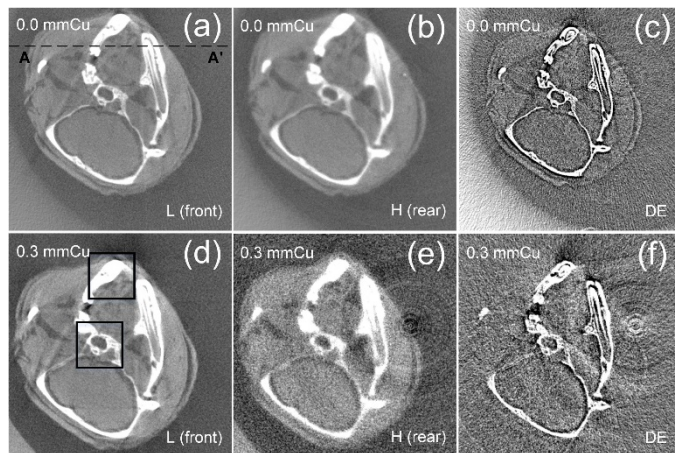
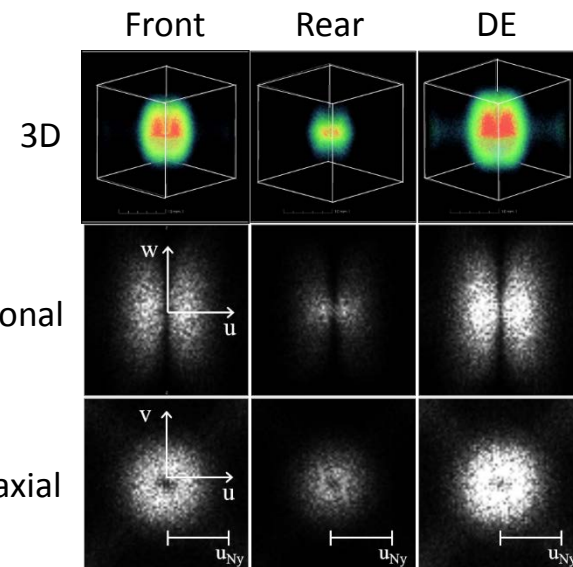
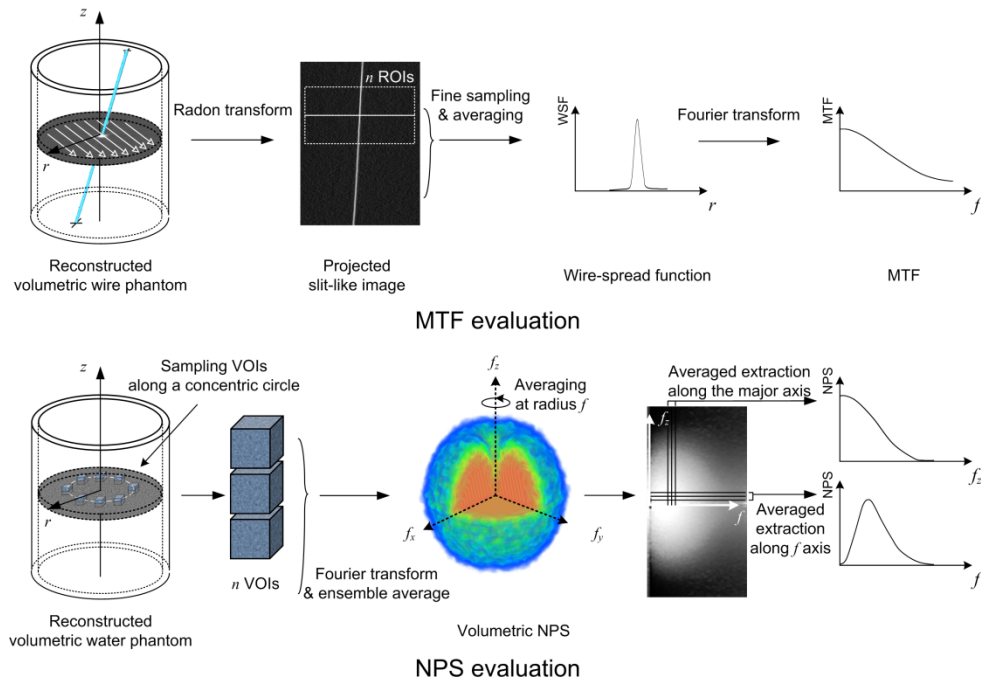


$$\text{NPS}_{DE}(\mathbf{u}) = \bar{w}^2 \text{NPS}_L(\mathbf{u}) + (1 + \text{SPR}) \text{NPS}_H(\mathbf{u})$$

$$\text{NEQ}_{DE}(\mathbf{u}) = \pi \mathbf{u} \frac{\text{MTF}_{DE}^2(\mathbf{u})}{\text{NPS}_{DE}(\mathbf{u})}$$

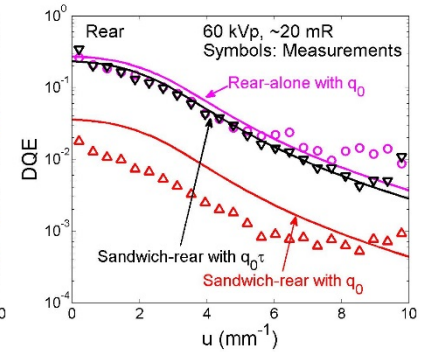
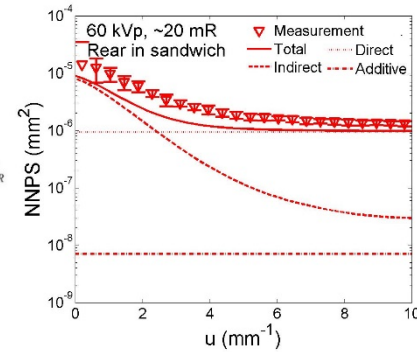
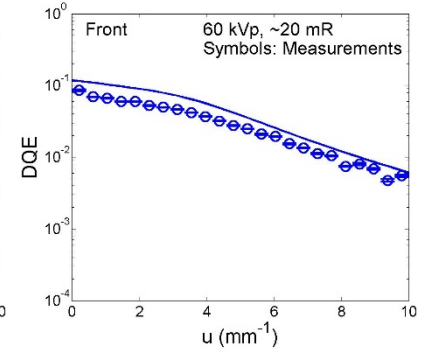
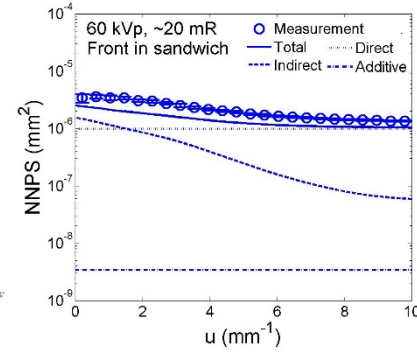
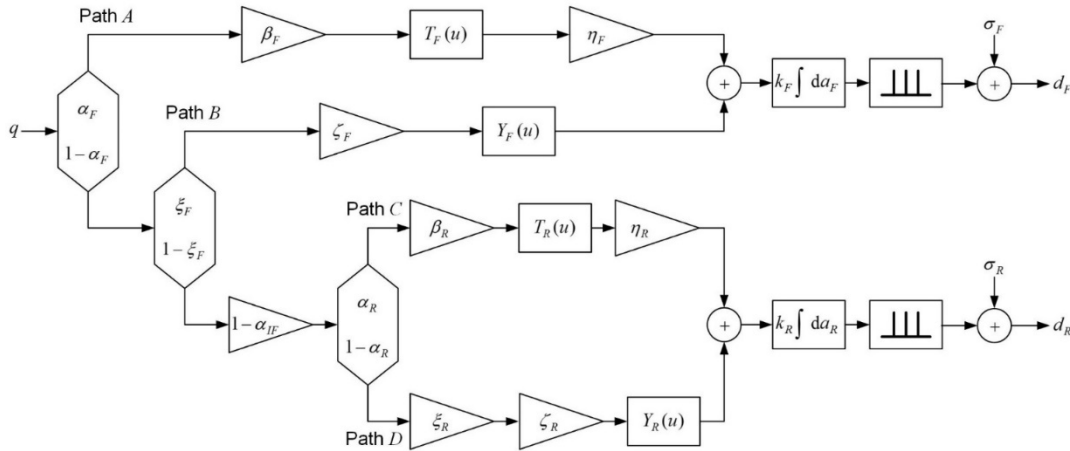
Imaging performance

S. Y. Jang *et al.* | IEEE TBME | in press | 2016



Manuscript in preparation

Analysis & optimization



$$FOM_k \approx (\Delta\mu_{kM}^R - w_{\hat{k}}\Delta\mu_{kM}^F)^2 t_k^2 \tau \left[\frac{w_{\hat{k}}^2 \tau}{A_{eff,F} DQE_F(0)} + \frac{1}{A_{eff,R} DQE_R(0)} \right]^{-1}$$

Wrap-up

- Unfortunately, detectors and systems are neither LSI nor stationary
- Nevertheless, the linear analysis (with reasonable assumptions) is useful to understand the working principle, and it can describe the actual performance in some limited extents
- The linear analysis can provides objective functions, $f(\boldsymbol{\theta})$, appropriate for optimizing design and technique parameters, $\boldsymbol{\theta}$

$$\boldsymbol{\theta}^* = \arg \max_{\boldsymbol{\theta}} f(\boldsymbol{\theta})$$

- Further consideration of nonlinear effects will result in better optimal parameters

Acknowledgements



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삼성전자



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INSTITUTE OF TECHNOLOGY



SAMSUNG MOBILE DISPLAY