# State of the art CT dose reduction and protocol optimization approaches for children

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Al related grant:

- CCRF ARC

Theranostics grants:

- CONNECT
- Clarity Pharmaceuticals





- Innovation in diagnostic CT technology and protocol development has led to significantly reduced pediatric patient dose levels
- This talk will discuss:
  - the key technologies that provide dose reduction
  - use of figure of merits (FOM) for image quality optimization
  - the effect of new, deep learning, CT reconstruction algorithms on image quality and patient dose reduction



#### **Historical Perspective**

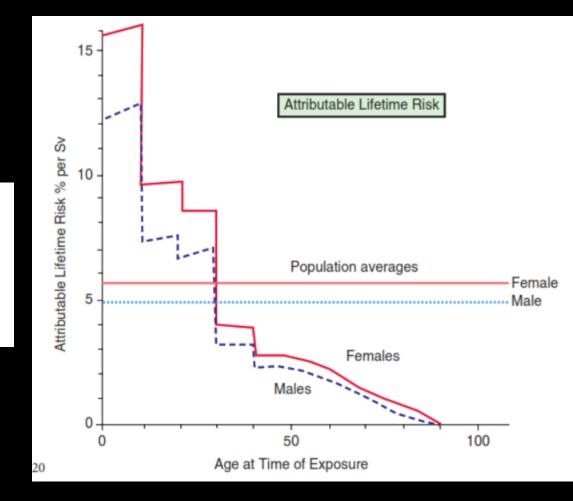
Pediatric sensitivity to radiation
It has been 20 years since Hall's paper

> Pediatr Radiol. 2002 Oct;32(10):700-6. doi: 10.1007/s00247-002-0774-8. Epub 2002 Jul 19.

Lessons we have learned from our children: cancer risks from diagnostic radiology

Eric J Hall 1

- Pediatric protocols have evolved
- Dose reduction has been significant





- #1 best dose reduction method:
  - Only scan when medically indicated!
- ACR appropriateness criteria
  - Physician/physics committee
  - Indication specific
    - Rank imaging modalities
    - Provides a dose estimate

| Relative<br>Radiation | Adult Effective<br>Dose Estimate | Pediatric<br>Effective Dose |
|-----------------------|----------------------------------|-----------------------------|
| Level*                | Range                            | Estimate Range              |
| 0                     | 0 mSv                            | 0 mSv                       |
| ۲                     | <0.1 mSv                         | <0.03 mSv                   |
| <b>*</b>              | 0.1 <b>-</b> 1 mSv               | 0.03-0.3 mSv                |
| €€€                   | 1-10 mSv                         | 0.3-3 mSv                   |
| €€€                   | 10-30 mSv                        | 3-10 mSv                    |
| ���€                  | 30-100 mSv                       | 10-30 mSv                   |

Variant 2: Left lower quadrant pain. Suspected complications of diverticulitis.

| Procedure  | Appropriateness Category | Relative Radiation Level |
|--|--------------------------|--------------------------|
| CT abdomen and pelvis with IV contrast                 | Usually Appropriate      | ***                      |
| CT abdomen and pelvis without IV contrast              | May Be Appropriate       | 888                      |
| CT pelvis with bladder contrast (CT cystography)       | May Be Appropriate       | ****                     |
| MRI abdomen and pelvis without and with IV<br>contrast | May Be Appropriate       | 0                        |
| Fluoroscopy contrast enema                             | May Be Appropriate       | ***                      |
| Fluoroscopy cystography                                | May Be Appropriate       | ***                      |
| MRI abdomen and pelvis without IV contrast             | May Be Appropriate       | 0                        |
| US abdomen transabdominal                              | May Be Appropriate       | 0                        |
| CT abdomen and pelvis without and with IV contrast     | Usually Not Appropriate  | ****                     |
| Radiography abdomen and pelvis                         | Usually Not Appropriate  | ***                      |
| US pelvis transvaginal                                 | Usually Not Appropriate  | 0                        |
|  |                          |                          |

- How is this applicable in RT?
  - It is not a question of if a patient needs a CT but how many
  - Do it right the first time: set the right dose!
    - Too low of dose usually equals non diagnostic exams which lead to repeat exams
    - Carefully position and double check settings



- How is this applicable in RT?
  - Follow up CTs
    - Is surveillance necessary?
    - How often should these occur?
    - Can follow up be reasonably be performed w/ MRI?

#### Is Routine Pelvic Surveillance Imaging Necessary in Patients With Wilms Tumor? Cancer January 1, 2013

Sue C. Kaste, DO<sup>1,2,3</sup>; Samuel L. Brady, PhD<sup>1</sup>; Brian Yee, RT(R)(CT), ARRT<sup>1</sup>; Valerie J. McPherson, BS<sup>2</sup>; Robert A. Kaufman, MD<sup>1,3,4</sup>; Catherine A. Billups, MS<sup>5</sup>; Najat C. Daw, MD<sup>6</sup>; and Alberto S. Pappo, MD<sup>2</sup>

90% cured Most relapses are salvaged All relapsed patients were symptomatic and didn't require CT surveillance for detection



Follow up CTs

**Pediatr Blood Cancer** 

The Role of Chest Computed Tomography (CT) as a Surveillance Tool in Children With High-Risk Neuroblastoma

Sara M. Federico, MD,<sup>1,4</sup>\* Samuel L. Brady, PhD,<sup>2</sup> Alberto Pappo, MD,<sup>1,4</sup> Jianrong Wu, PhD,<sup>3</sup> Shenghua Mao, PhD,<sup>3</sup> Valerie J. McPherson, BS,<sup>1</sup> Alison Young, MSFNP,<sup>1</sup> Wayne L. Furman, MD,<sup>1,4</sup> Robert Kaufman, MD,<sup>2,4,5</sup> and Sue Kaste, DO<sup>1,2,5</sup>

Relapsed patients only have a 10% salvage rate Patients' w/o thoracic disease at diagnosis were otherwise symptomatic

Reduction or removal of chest CT for surveillance leads to 35-40% dose reduction



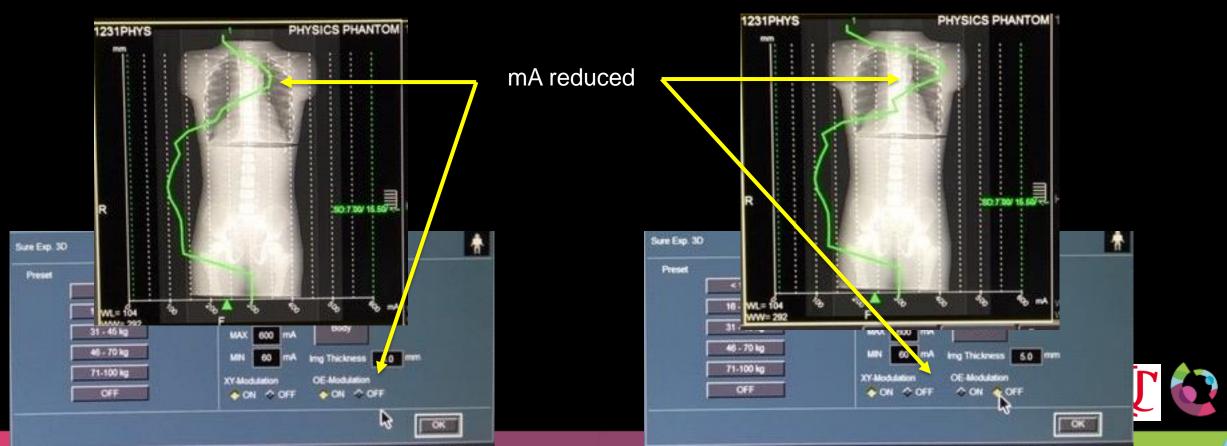
#### **Tube Current Modulation**

The #2 best dose reduction methodology in CT

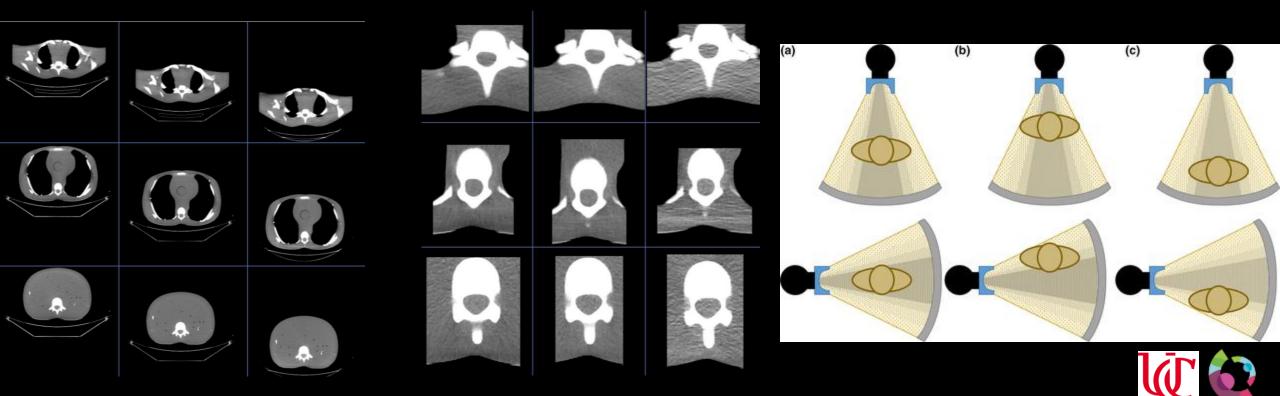


#### Organ Dose Modulation

- Organ dose modulation reduces mA over anterior portion of the body
  - Used to reduce eye lens, thyroid, and breast dose
  - Used along with TCM for additional dose reduction NO IQ PENALTY



- Affects patient dose when using TCM
  - Patients lower in the gantry lead to beam hardening and photon starvation artifacts

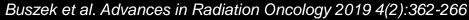


Images belong to Timothy Szczykutowicz, PhD

Barreto et al. JACMP 2019 20(6); 141-151

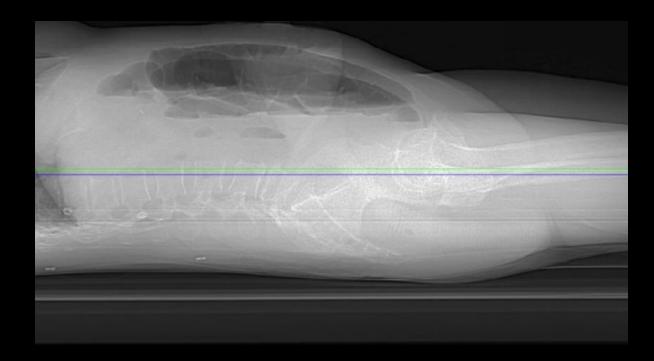
- How is this applicable in RT?
  - May be difficult to find a patient's actual center
    - Due to immobilization devices
    - OR nontypical supine positions





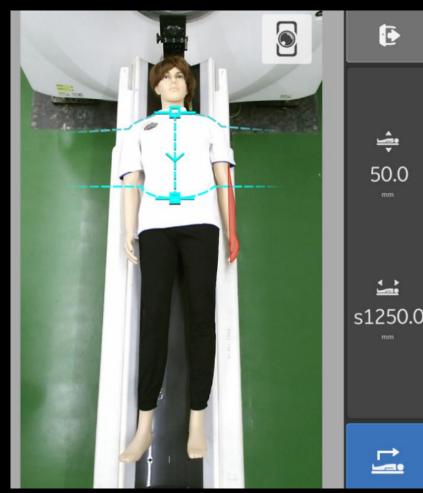


- Patient centering verification based on attenuation map
  - Some CT vendors will allow a single click move to center
  - Others require technologists' manual intervention





- Al algorithms + camera
  - Visual light based cameras identify anatomical landmarks
  - Provide centering & position guidance
  - Consistent scan coverage
    - Appropriate scan coverage helps reduce unnecessary patient dose

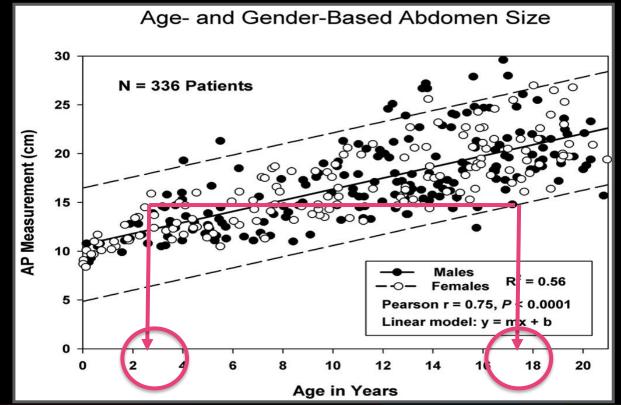


GE AI-based Auto Positioning, white paper



#### **Patient Size**

- Same age patients vary dramatically in size
  - Abdomen of smallest 17-yr-old and largest 3-year-old are same size
  - Use patient cross sectional size not age or even weight when setting protocols





#### **Patient Size**

- Use of scan projection radiograph (SPR) to set protocols
  - Measure patient attenuation or "size"
  - Protocol selected based on measured size
- Use of SSDE to better approximate patient dose

- SSDE is calculated for body (TG 204 & 220) and the head (TG 293)

 $SSDE = CTDI_{vol} \cdot f_{size}$ 

– Where  $f_{size}$  can be found by measuring patient attenuation or effective dia.

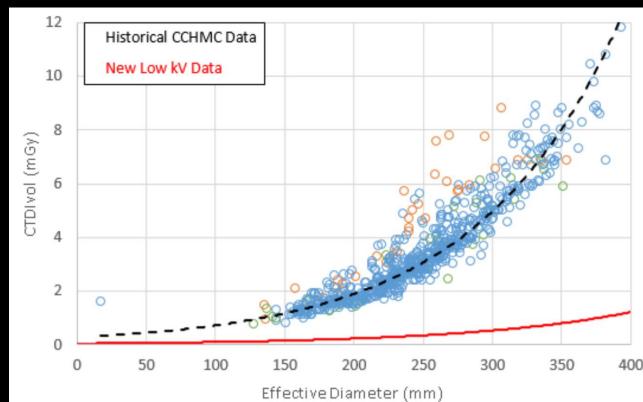
- Why do we use low kV imaging in Radiology?
  - Lower radiation dose to the patient
  - Better tissue <u>contrast</u> differentiation
  - Why does the dose go down?
    - Dose increases/decreases linearly with mA, but quadratically with kV
  - If we reduced kV by 40:  $\left(\frac{80 \ kV}{120 \ kV}\right)^2$  we get 56% dose reduction
  - If we reduce mA by 40:  $\left(\frac{80 \ mA}{120 \ mA}\right)$  we get 33% dose reduction

#### Remember:

Changing mA affects dose/noise Changing kV affects dose/noise, image contrast, AND CT #



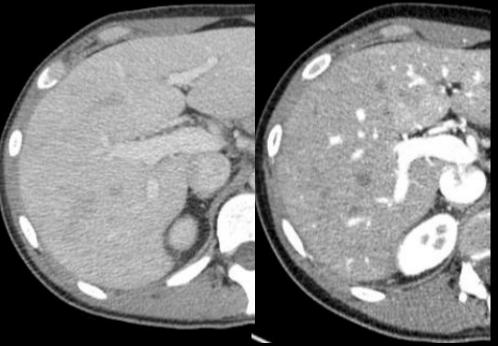
- Lower kV requires high mAs capacity
  - Modern scanners use high tube current (e.g., 1200 mA) w/ low kV
  - Deliver lower dose for all patient sizes

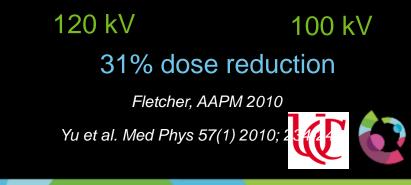




- Lower kV requires more mAs for similar exposure/noise
  - What is the correct mAs/eff mAs
- Don't match noise, match CNR!
  - CNR improves with lower kV even though noise increases
  - Noise may be higher at lower kV than at 120 kV

#### Same Patient





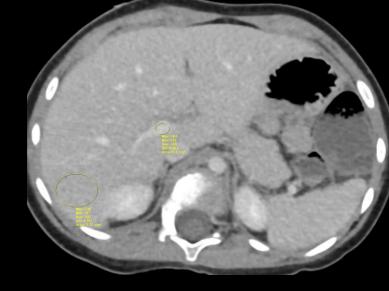
- Rule of thumb
  - Routine body imaging @ 70 kV
    - < 30 cm AP+LAT (CCHMC)
    - Typically, neonates (< 15 kg)</li>
  - Routine body imaging @ 80 kV
    - 30-60 cm AP+LAT (CCHMC)
    - Typically, toddlers to large teenagers
  - Routine body imaging @ 100 kV
    - > 60 cm AP+LAT (CCHMC)
    - Typically, large teenagers young adults

- Rule of thumb
  - Routine imaging @ 100 kV
    - Heads < 5 years old
  - Routine imaging @ 120 kV
    - Heads > 5 years old

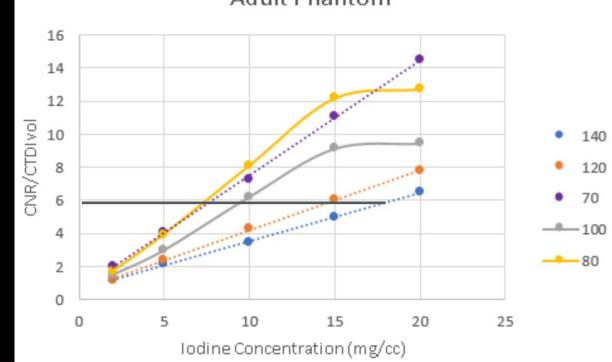


- Lower kV protocols may lead to lower IV contrast
  - This is largely true for all pediatric patients
  - Limited for adults





Adult Phantom

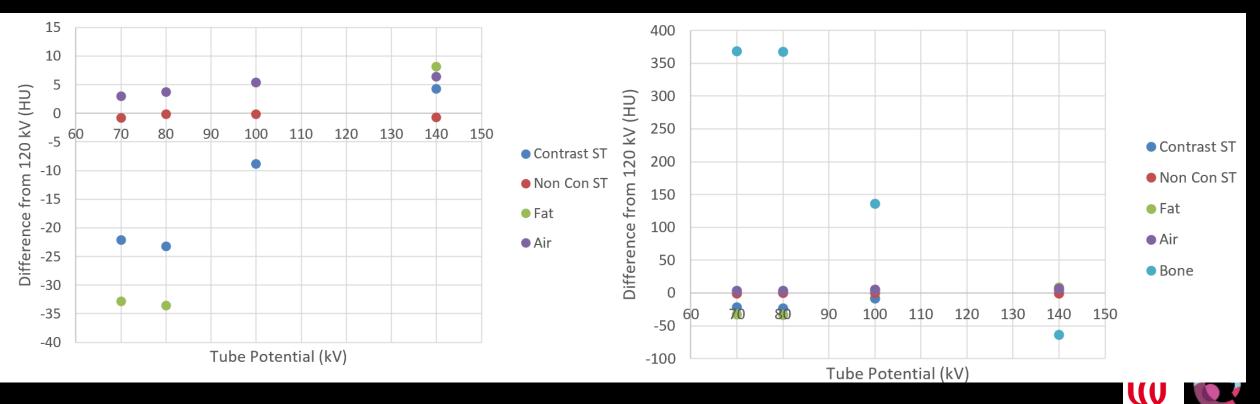


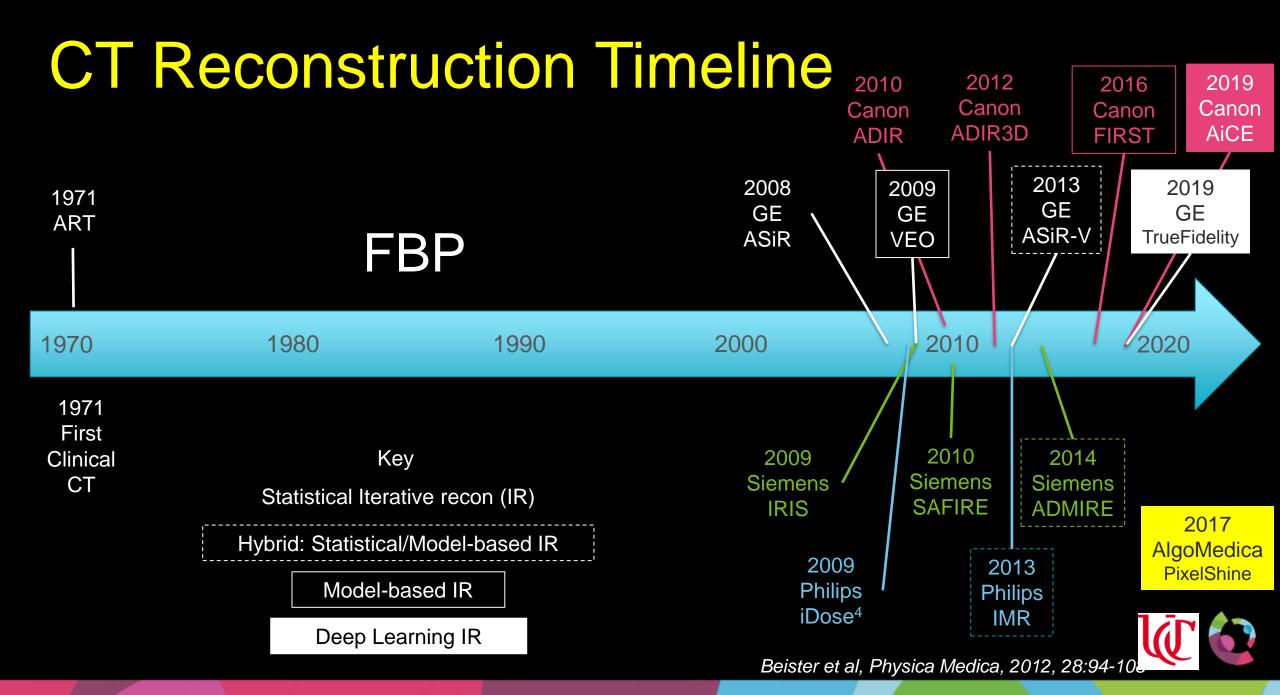
- How is this applicable in RT?
  - When changing kV, CT number changes for high attenuating material
    - Bone & contrast infused tissues
  - Less change for soft tissue
    - Water and Air CT #s stay the same for all kV's

$$\mathrm{HU} = \left(\frac{\mu_{\mathrm{material}} - \mu_{\mathrm{water}}}{\mu_{\mathrm{water}}}\right) \times 1000$$

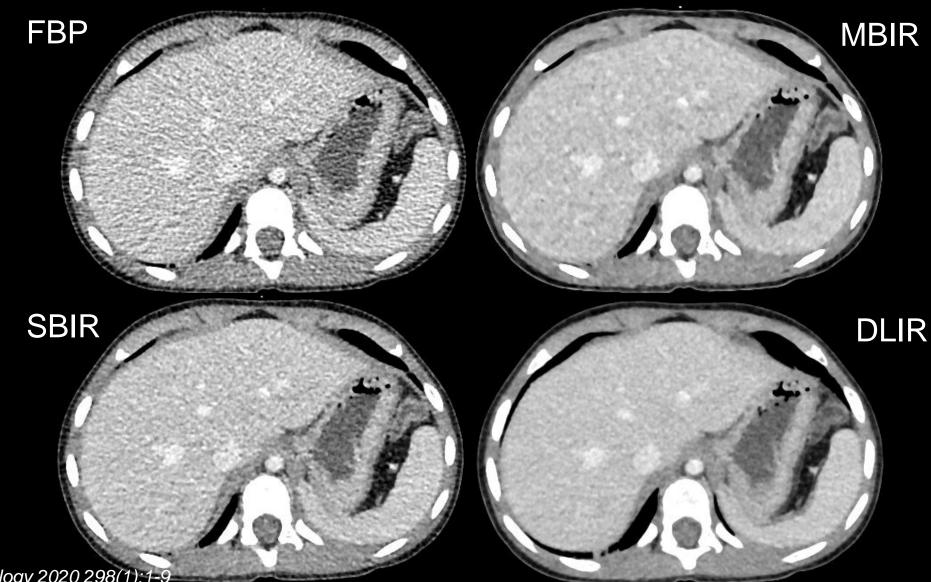
- Minimize CT # changes to keep changes in Tx dose by <1%\*</li>
  - Changes in soft tissue #s are more detrimental to Tx doses than to bone
  - Suggest that CT # changes be kept to:
    - $\pm$  20 HU for soft tissue &
    - $\pm$  50 HU for the lung and bone

- Suggest that CT # changes be kept to:
  - ± 20 HU for soft tissue &
  - ± 50 HU for the lung and bone





#### **Image Reconstruction-Options**



Brady et al. Radiology 2020 298(1):1-9



#### **CT DLIR-CCHMC Experience**

- How is this applicable in RT?
  - No measurable difference in CT # between DLIR and IR
  - Noise reduction, CNR improvement do not affect dose planning accuracy
  - Additionally: shown to improve organ segmentation time/accuracy

**>** J Med Imaging (Bellingham). 2019 Oct;6(4):043504. doi: 10.1117/1.JMI.6.4.043504. Epub 2019 Oct 24.

Deep learning-based image quality improvement for low-dose computed tomography simulation in radiation therapy

Tonghe Wang <sup>1</sup>, Yang Lei <sup>1</sup>, Zhen Tian <sup>1</sup>, Xue Dong <sup>1</sup>, Yingzi Liu <sup>1</sup>, Xiaojun Jiang <sup>1</sup>, Walter J Curran <sup>1</sup>, Tian Liu <sup>1</sup>, Hui-Kuo Shu <sup>1</sup>, Xiaofeng Yang <sup>1</sup>

**REVIEW** article

Front. Oncol., 01 October 2019 Sec.Radiation Oncology https://doi.org/10.3389/fonc.2019.00977

#### Deep Learning: A Review for the Radiation Oncologist



Carlotta Masciocchi<sup>1</sup>,

Yanting Shen<sup>3</sup>

#### **CT DLIR-CCHMC Experience**

- Two vendors, two installs, two years apart
  - Canon's AiCE installed on Aquilion One Genesis
  - GE's TrueFidelity installed on Revolution Apex
- Two different approaches to implement DLIR
  - AiCE install occurred first (2019)
  - We needed to sort through all the reconstruction options
    - Learn radiologist preference(s)
    - Test for diagnostic confidence

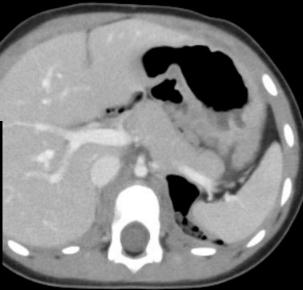


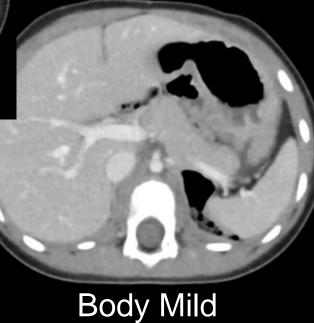
Canon's AiCE



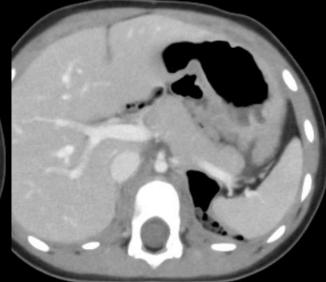


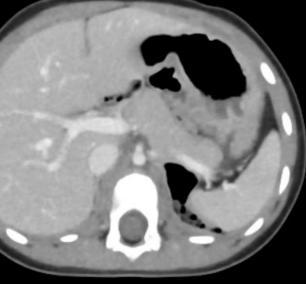
#### Body Sharp Mild



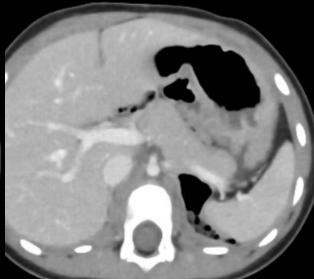


#### Body Sharp Standard Body Sharp Strong

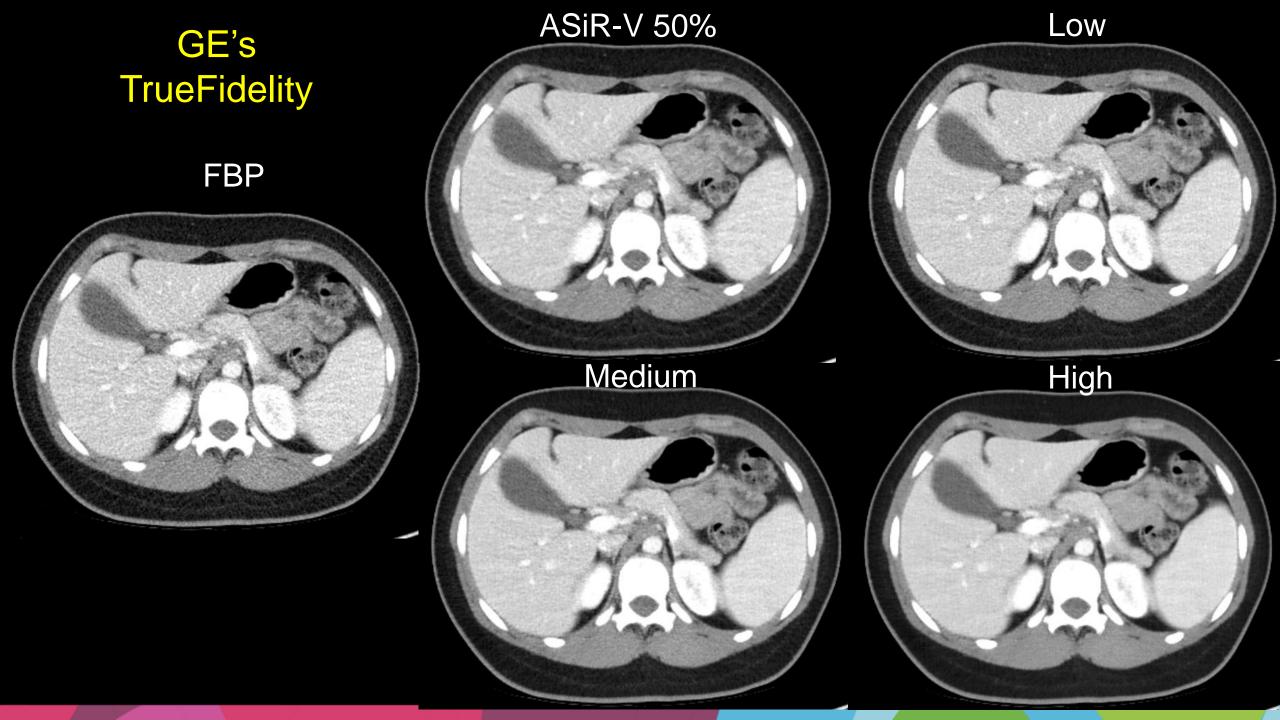


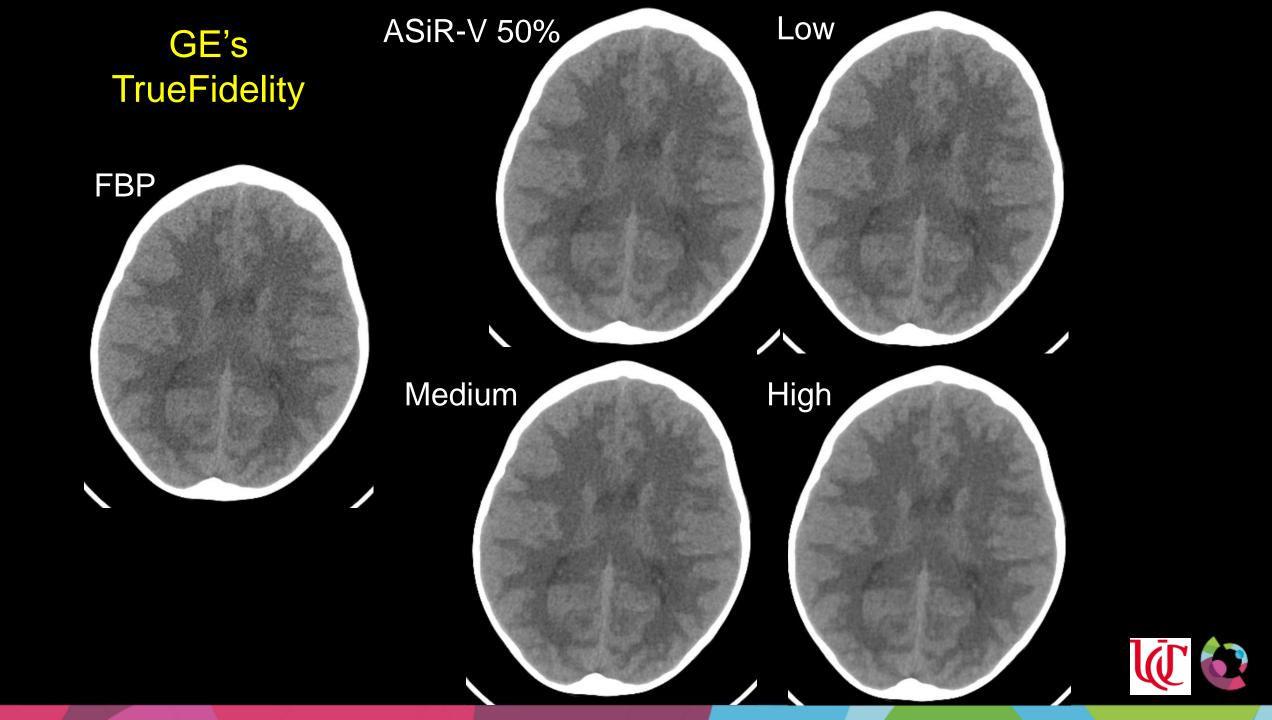


**Body Standard** 



Body Strong





- To learn radiologist preference(s) & test for diagnostic confidence
  - We selected a variety of patient ages/sizes for reconstruction
    - Total was ~130 exams
      - Each patient was reconstructed using clinical SBIR + 6 DLIR options
  - Each exam was evaluated
    - 1. Mathematical observer/rater [using a non-prewhitening-matched mathematical-observer model with eye filter  $(d'_{NPWE})$ ]
    - 2. Human observer/rater
    - 3. Took all data and did an ROC analysis



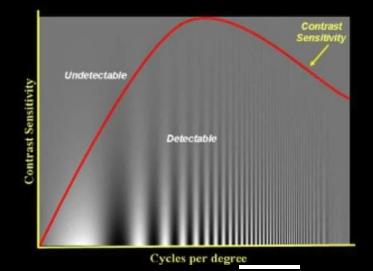
 Objective model: non-prewhitening matched detection index observer model (d'<sub>NPWE</sub>):

$$d'_{NPWEi} = \frac{\left[2\pi \cdot \int_{0}^{Nyquist} |W(f)|^{2} \cdot TTF(f)^{2} \cdot E(f)^{2} \cdot df\right]^{2}}{2\pi \cdot \int_{0}^{Nyquist} [|W(f)|^{2} \cdot TTF(f)^{2} \cdot E(f)^{4} \cdot NPS(f) + N(f) \cdot NPS(f)] df}$$

- Eye Function: models the eye response to spatial freq

$$-E(f) = \rho^{1.5} \cdot e^{-C \cdot \rho^2}; \ \rho = f \cdot \frac{\pi}{180} \cdot \frac{d_v \cdot SFOV}{DFOV(cm)}$$

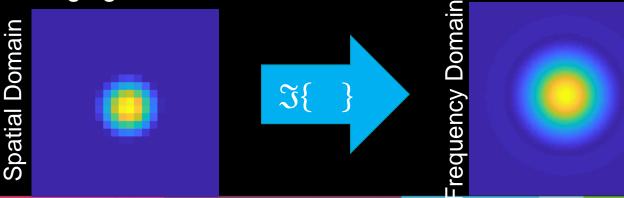
\*Richard S, Siewerdsen JH. Med Phys. 2008;35(11):5043-53. \*Solomon J, Samei E. J Med Imaging (Bellingham). 2016;3(3):035506.



 Objective model: non-prewhitening matched detection index observer model (d'<sub>NPWE</sub>):

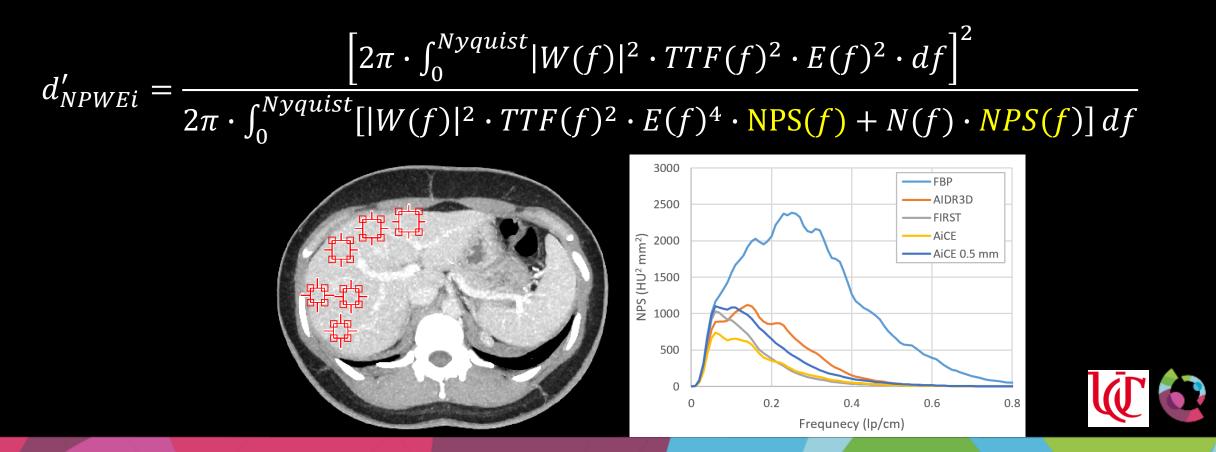
$$d'_{NPWEi} = \frac{\left[2\pi \cdot \int_{0}^{Nyquist} |W(f)|^{2} \cdot TTF(f)^{2} \cdot E(f)^{2} \cdot df\right]^{2}}{2\pi \cdot \int_{0}^{Nyquist} [|W(f)|^{2} \cdot TTF(f)^{2} \cdot E(f)^{4} \cdot NPS(f) + N(f) \cdot NPS(f)] df}$$

- -W(f) is the task function, i.e., the Fourier transform of the signal to be detected
  - Circular objects ranging from 0.5 to 10 mm





- Noise Power Spectrum [NPS(f)]
  - Patient images imported into IMQUEST, Duke University



 Objective model: non-prewhitening matched detection index observer model (d'<sub>NPWE</sub>):

$$d'_{NPWEi} = \frac{\left[2\pi \cdot \int_0^{Nyquist} |W(f)|^2 \cdot TTF(f)^2 \cdot E(f)^2 \cdot df\right]^2}{2\pi \cdot \int_0^{Nyquist} [|W(f)|^2 \cdot TTF(f)^2 \cdot E(f)^4 \cdot NPS(f) + N(f) \cdot NPS(f)] df}$$

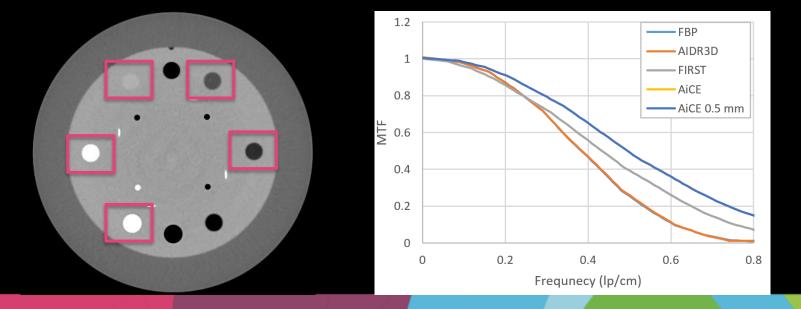
- -N(f) is a scalar to model the human inefficiency caused by cognitive inconsistency
  - Defined as 60% of NPS based on prior studies\*





- Task Transfer Function [TTF(f)]
  - CatPhan 600 Phantom imported into IMQUEST, Duke University

 $d'_{NPWEi} = \frac{\left[2\pi \cdot \int_{0}^{Nyquist} |W(f)|^{2} \cdot TTF(f)^{2} \cdot E(f)^{2} \cdot df\right]^{2}}{2\pi \cdot \int_{0}^{Nyquist} [|W(f)|^{2} \cdot TTF(f)^{2} \cdot E(f)^{4} \cdot NPS(f) + N(f) \cdot NPS(f)] df}$ 



- Non-prewhitening matched detection index observer model (d'<sub>NPWE</sub>)
   Used as a metric of SNR
- Use  $d'_{NPWE}$  to calculate an area under the curve  $(A_z)$  score

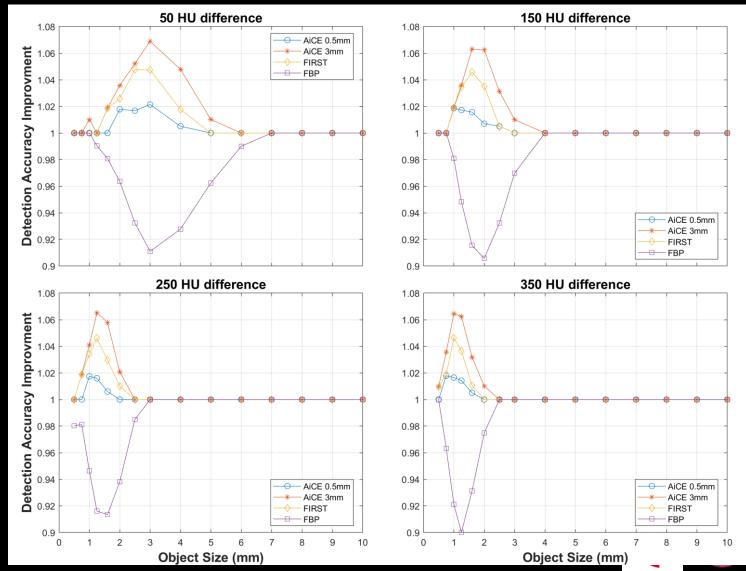
$$A_{z} = \frac{1}{2} \cdot \left[ 1 + \frac{2}{\sqrt{\pi}} \cdot \int_{0}^{d'_{NPWEi}/2} e^{-x^{2}} dx \right]$$

- Used as a metric for detection accuracy
  - Calculated at each object size (0.5 to 10 mm)
  - Calculated at each contrast difference level (50 to 350 HU; increments of 100 HU)



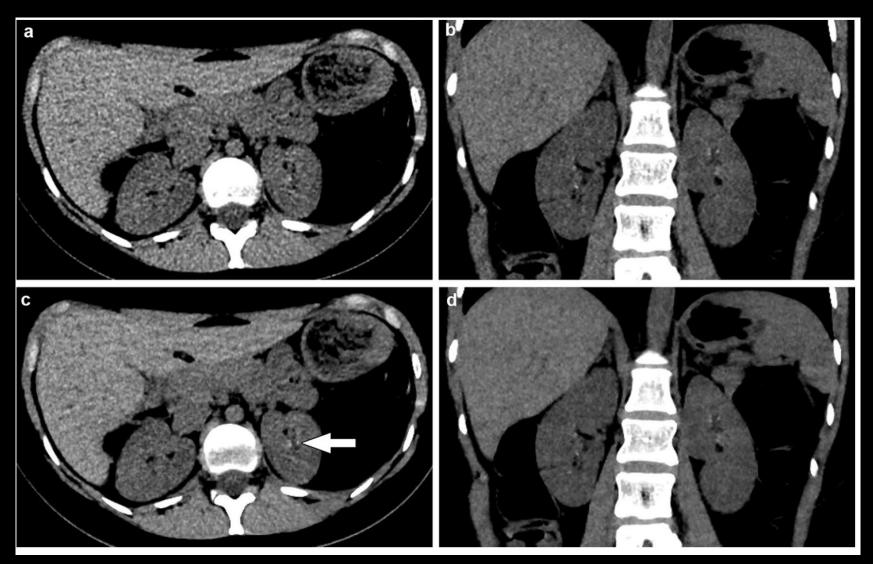
#### Mathematical Evaluation

- Mathematical observer study
  - AUC scores normalized to AIDR3D (SBIR)
    - 5mm thick images
  - AUC scores averaged over all patients
  - Results:
    - DLIR > MBIR > SBIR
    - DLIR(0.5 mm) > SBIR(5 mm)



Brady et al. Radiology 2020 298(1); doi.org/10.1148/radiol.2020202317

#### **Object Detection**

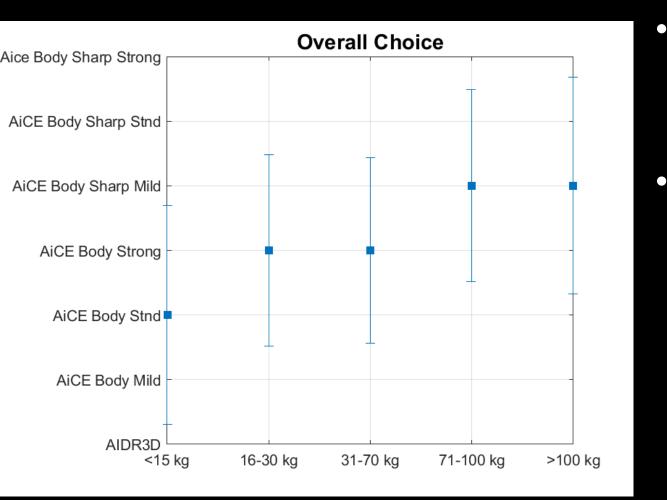


22-year-old woman with flank pain. Axial and coronal reformatted images reconstructed with iterative reconstruction (a and b) and deep learning reconstruction (c and d) show multiple calculi in both kidneys. One of the two stones in the upper pole of the left kidney is not visible in the axial image reconstructed using iterative reconstruction (a) and is visible on the deep learning reconstruction image (white arrow in c). All stones are visible on both reconstructions in the coronal plane. Note the decreased image noise in the deep learning reconstruction image



Thapaliya et al. Abdom Radiol 2022 47(1):265-271; doi.org/10.1007/s00261-021-03274-7

#### **Observer Preference**



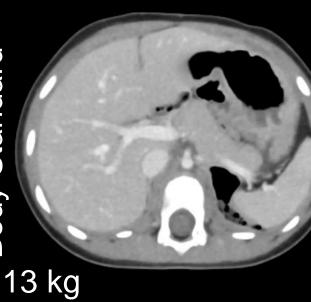
- Is there a preference for the use of DLIR by patient size/weight?
- When considering all aspects of the image, in a blinded observer study, the participants demonstrated DLIR preference by patient weight



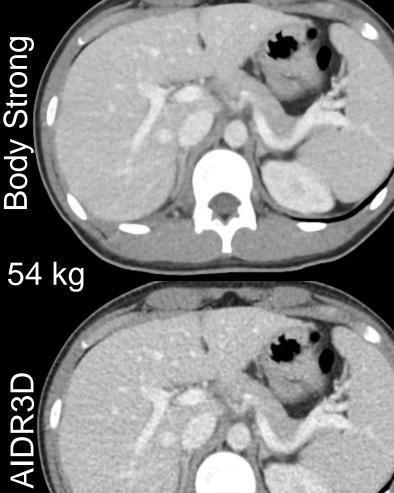




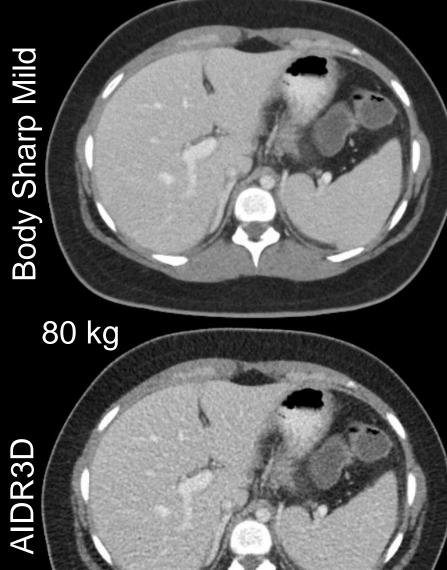
# **Body Standard**



# **Body Strong**



## AIDR3D



#### AiCE-Body Standard



#### TrueFidelity-Medium



#### Conclusions

- Use tube current modulation!
  - Use organ dose modulation if available
- Center your patients
- Create size-based protocol
- Reduce kV where possible (be careful of CT # change)
  - Dose reduction & CNR improvement
  - Potential reduction of IV/oral contrast concentration
- Adopt DLIR when possible



#### Thank you



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