Next Generation Radiotherapy Treatment Planning Based on GPU and Cloud Computing

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GPU Graphics Processing Unit



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4th Short Course on GPU Programming for Medical Physics and Medical Imaging Research

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Sponsor: Center for Advanced Radiotherapy Technologies (CART), University of California San Diego (UCSD), La Jolla, CA 92093 ased ultrafast IMRT plan optimization

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Place: San Diego Supercomputer Center (SDSC), University of California San Diego

Course Director:

Steve Jiang, Ph.D., Professor, CART, UCSD Course Co-directors: Xun Jia, Ph.D., Assistant Professor, CART, UCSD Amitava Majumdar, Ph.D., Associate Professor, SDSC and CART, UCSD

GPU-based ultra-fast direct aperture optimization for online adaptive radiation therapy

Chunhua Men, Xun Jia and Steve B Jiang Department of Radiation Oneslogi, University of California San Diego, La Jolis, CA 92027, USA E-mail: styling@cocdad Received 28 March 2010, in final form 17 June 2010 Published 20 July 2010

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Medical Physics Letter

Ultrafast treatment plan optimization for volumetric modulated arc therapy (VMAT)

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NBC Evening News

Dose Engine #1: GPU-based FSPB model

U Version 1

- Gu et al Phys Med Biol. 54(20):6287-6297, 2009
- Conventional FSPB model

Version 2

- Gu et al Phys Med Biol. 56(11): 3337-3350, 2011
- With 3D density correction
- Accuracy greatly improved
- Still extremely efficient: <1 s for IMRT, ~15 s for VMAT

Dose Engine #2: GPU version of DPM MC code

Version 1

- Jia et al Phys Med Biol. 55(11): 3077–3086, 2010
- A straightforward implementation
- Only 5-6x speed up
- Version 2
 - Jia et al Phys Med Biol. 56(22):7017-7031, 2011
 - More GPU friendly
 - ~ 60x speed up
 - < 40 s for IMRT and VMAT
- Version 3
 - Townson et al Phys Med Biol. 58(12):4341-4356, 2013
 - Phase space files and commissioning procedure







Dose Engine #3: gPMC – GPU-based Proton MC













gPMC

Phantom	Energy (MeV)	3%/3mm Gamma passing rate (%)	Time (sec)	CPU time (hr)	
Water	100	100	6.7	10.8	
Water	200	100	22.1	39.5	
Bone	100	100	6.1	20.8	
Bone	200	99.9	19.6	70.0	
Plastic	100	100	6.3	10.2	
Plastic	200	99.9	20.8	78.7	
Inhomogeneous	100	100	8.6	15.9	

Paganetti, Jiang, Phys Med Jia, Schuemann, Paganetti, Ji *Biol*, 57(23):7783-7797, 2012 10

Three GPU-based Optimization Models



Men et al Phys Med Biol. 54(21):6565-6573, 2009

< 1 second

Men et al Phys Med Biol. 55(15):4309-4319, 2010





~ 2 seconds

Men et al *Med Phys* 37(11): 5787-5791, 2010 Peng et al *Phys Med Biol*. 57(14):4569-88, 2012

~ 20-60 seconds

Deformable Image Registration (DIR) for ART

Demons on GPU

• Gu et al *Phys Med Biol* 55(1): 207-219, 2010

Contour-guided DIR

• Gu et al *Phys Med Biol* 58(6):1889-1901, 2013

CT/CBCT DIR with intensity correction

Zhen et al Phys Med Biol 57: 6807-6826, 2012

CT/CBCT DIR with truncation

Zhen et al Phys Med Biol 58:7979-7993, 2013

1. Implement Demons DIR on GPU

Gu, Pan, Liang, Castillo, Yang, Choi, Castillo, Majumdar, Guerrero, and Jiang *Phys Med Biol 55(1): 207-219, 2010*

GPU-based Demons DIR

Iterative method

$$\mathbf{u}^{k+1} = \left(\mathbf{u}^{k} + \frac{\left(I(\mathbf{x} + \mathbf{u}^{k}) - J(\mathbf{x})\right)\nabla J(\mathbf{x})}{\left|\nabla J(\mathbf{x})\right|^{2} + \left(I(\mathbf{x} + \mathbf{u}^{k}) - J(\mathbf{x})\right)^{2}}\right) \circ G(\sigma)$$

> Key assumptions:

Intensity constancy: a point in the Moving image I(x)
 looks the same in the Target image J(x)
 Small motion

> Six variants:

- 1. Passive force; 2. Evolved Passive force; 3. Active force;
- 4. Double force; 5. Normalized double force; 6. Inverse consistency;

GPU-based Demons DIR

- > Images size ~256*256*100
- > ~100x speedup compared to an Intel Xeon 2.27 GHz CPU

Computational time (unit: sec)							
Method	Case 1	Case 2	Case 3	Case 4	Case 5	Average	
1	6.80	7.18	7.39	6.49	7.24	7.02	
2	6.82	7.20	7.42	6.56	7.08	7.02	
3	8.29	9.24	8.79	7.75	8.44	8.50	
4	7.71	8.65	8.02	8.30	8.44	8.22	
5	8.36	8.69	8.97	7.77	8.70	8.50	
6	11.07	11.47	11.54	10.46	10.98	11.10	

2. Contour-guided Demons DIR

- After DIR contour propagation, manual inspection and revision (if needed)
- DVF updating for accurate dose accumulation

Gu, Dong, Wang, Yordy, Mell, Jia and Jiang *Phys Med Biol.* 58(6):1889-1901, 2013.

Contour-Guided Deformable Registration

Energy function:

$$E(d\mathbf{r}) = \frac{1}{2} \left\| I_0(\mathbf{r} + d\mathbf{r}) - T_0(\mathbf{r}) \right\|_2^2 + \sum_i \frac{\lambda_i}{2} \left\| I_{s_i}(\mathbf{r} + d\mathbf{r}) - T_{s_i}(\mathbf{r}) \right\|_2^2$$
$$d\mathbf{r} = \arg\min E(d\mathbf{r})$$

Key assumption:

$$d\mathbf{r} = v\mathbf{n}$$
, with $\mathbf{n} = \frac{\nabla I_0(\mathbf{r}) + \sum_i \lambda_i \nabla I_{s_i}(\mathbf{r})}{|\nabla I_0(\mathbf{r}) + \sum_i \lambda_i \nabla I_{s_i}(\mathbf{r})|}$
Solution:

$$v = \frac{(\nabla I_0(\mathbf{r}) \cdot \mathbf{n})(T_0(\mathbf{r}) - I_0(\mathbf{r})) + \sum_i \lambda_i (\nabla I_{s_i}(\mathbf{r}) \cdot \mathbf{n})(T_{s_i}(\mathbf{r}) - I_{s_i}(\mathbf{r}))}{(\nabla I_0(\mathbf{r}) \cdot \mathbf{n})^2 + \sum_i \lambda_i (\nabla I_{s_i}(\mathbf{r}) \cdot \mathbf{n})^2 + (T_0(\mathbf{r}) - I_0(\mathbf{r}))^2}$$

Contour-Guided Deformable Registration



Red lines: manual contours on pCT and mapped on tCT through rigid registration Blue lines: manual contours on tCT Yellow line: deformed contours through demons registration Green line: deformed contours through CG-DIR registration

3. CT/CBCT DIR with Intensity Correction

Zhen, Gu, Yan, Zhou, Jia, and Jiang Phys Med Biol 57: 6807-6826, 2012

Intensity Inconsistency between CT and CBCT

- □ Scatter artifacts in CBCT
- **D** Bowtie filter artifact
- **Different scan geometry**
- **Different level of noise, beam hardening, etc**









4. CT/CBCT DIR with truncation

Zhen, Yan, Zhou, Jia, Jiang, *Phys Med Biol*. 58(22):7979-7993, 2013.

CBCT Image Truncation

Causes

- -Sometimes we only scan ROI



Problems in CT/CBCT DIR

- -Unphysical results in/near the missing regions
- Errors in dose calculation for ART

Rationale of Our Algorithm





Information is missing in image domain But it still exists in projection domain

A Hybrid Reconstruction/Registration Algorithm

$$\arg\min E(x) = \frac{1}{2} \left\| Pf^{0}(x+v(x)) - g \right\|_{2}^{2} + \frac{\lambda}{2} \left\| \nabla v \right\|_{2}^{2}$$

•*P*: projection operator •*v*: vector

• f⁰: moving image(CT) • g: CBCT projection

v: vector field

Euler–Lagarange equation:









Truncated CBCT



Deformed CT





Truncated CBCT proj

Our Method

Demons



Deformed CT

СТ

The Use of GPU Tools

Improve efficiency for some computational tasks in radiotherapy

Over the second seco

Online Adaptive Radiotherapy



Tumor volume shrinkage in response to the treatment

- Tumor shape deformation due to filling state change of neighboring organs
- Relative position change between tumor and normal organs
- Beyond geometrical changes

SCORE <u>Supercomputing On-line Re-planning Environment</u>



A GPU-based real-time automatic re-planning system A research platform for online and offline ART Clinical studies: H/N, pancreas, GYN, prostate, lung, *etc*

H & N Case 1

Planning CT

Treatment CT



7 beams at 0, 51, 102, 153, 207, 258 and 309 degree angles



Original plan vs. Un-adapted plan





Online Re-planning - A Paradigm Shift

Past and current: <u>plan-centered</u>

- A snapshot of patient anatomy before treatment
- A treatment plan based on this snapshot
- Try to match this plan with the patient anatomy throughout the whole treatment course

Generation: Future: patient-centered

- Automatic plan re-optimization on CBCT every day
- Setup errors and anatomical variations are considered in the new plan
- Much smaller PTV margin and faster patient setup

From Adaptive Therapy Re-planning To Next Generation Treatment Planning

Problems with Current Treatment Planning



Large variation in plan quality (on paper and delivered)

- Inter- and intra-institution variation
- Inter- and intra-TPS variation
- Inter- and intra-planner variation
- Inter- and intra-attending variation

Das IJ et al, J Natl Cancer Inst. 100(5):300-7, 2008.

Problems with Current Planning - Low Efficiency



The whole process may take a week !

Problems with Current Planning – High Cost



Large portion of healthcare professionals' time (Time is money!)

Planning system purchase, installation, commissioning, maintenance, and upgrade





A dedicated data center is needed for patient data storage

Problems with Current Treatment Planning

- **Difficult for cross institution collaboration**
- **Difficult for experience sharing**
- Difficult for data sharing and mining
- Difficult for multi-institution clinical trials

Our Vision on Next Generation TP

□ Based on recent technologies

—GPUs, cloud computing, and iPad

Physician centered and a two step approach

□ 1st step: automatic planning (Cloud computing)

- Using delivered plans of previously treated patients as reference
- The auto plan should be good and robust enough for clinical use for most standard cases
- **2**nd step: interactive tuning (GPU and iPad)
 - Physician interactively tune the plan by dragging DVH and iso-dose curves
 - Plan re-optimization in real-time in response to any change

Proposed Process for Next Generation TP

- **Upload patient CT and treatment protocol to the GPU cloud**
- Automated contouring and planning in the cloud
- Clinician to review and interactively revise the contours and plan using an iPad
- Download the plan, QA, and treat





Benefits for Next Generation TP

□ Higher plan quality

- Reduced variation
- Important for resource-limited institutions/regions
- Lower cost
 - Reduce costs in purchasing, installing, commissioning, maintaining, and upgrading TPS
 - Improved resource utilization ratio
- □ Higher efficiency
 - Better utilization of professional's time
 - High throughput
- Better usability and accessibility
 - Planning can be done anywhere on any device
- □ Faster deployment of new technologies
- Enhanced collaborative clinical trials and data mining

Optimization Guided by A Reference Plan

- Prior knowledge based optimization
- **Guided with reference DVHs**
 - Clinician approved trade-offs among PTVs/OARs
 - Clinician approved trade-offs among dose-volume criteria for each PTV/OAR

Guided with reference dose distribution

- Locations of hot/cold spot
- Shapes of iso-dose curves

Zarepisheh et al, *Phys Med Biol.* 58(6):1869-1887, 2013. Zarepisheh et al, *Med. Phys.* (accepted)

Where Are We?

Automatic contouring

Many commercial solutions

Automatic treatment planning

- Model based approaches (predicting DVH)
- Library based approaches (selecting reference patients)
 - Most similar patient as reference
 - Many patients as references

□ Interactive plan tuning

- Approximate approaches (cutting corners)
- GPU-based approaches (exact)

Demonstration of Some Preliminary Work

GPU-based brute force automatic planning

- A library of 100 prostate IMRT patients
- Leave-one-out test
- 99 new plans were automatically generated for each test patient in 80 minutes on one GPU card
- A GUI for physician to select a preferred plan
- Video (<u>YouTube</u>) (<u>Local</u>)

SCORE for automatic planning and interactive tuning

- SCORE was designed for ART re-planning
- It is being extended for automatic planning and interactive tuning
- A previously treated patient is chosen as the reference patient
- A plan for the new patient is automatically developed and then interactively tuned
- Video (<u>YouTube</u>) (<u>Local</u>)