CBCT images are required. Since simple double gating yields severe sparseness artifacts we propose a 5D motion compensation (MoCo) algorithm dedicated to cardiorespiratory CBCT in IGRT.

Material and Methods: Clinical patient data acquired with the TrueBeam™ CBCT system (Varian Medical Systems, Palo Alto, CA) have been used for our study. For the intrinsic respiratory and cardiac motion signal detection, about hundred overlapping regions of interests are automatically evaluated in projection space, thus yielding a robust approach independent on the anatomy shown in the projection images. In addition to respiratory gating (4D CBCT) cardiac gating is applied to obtain initial volumes. We compensate respiratory and cardiac motion in a two-step procedure. First, respiratory motion is estimated and compensated using respiratory phase binning only. Then, cardiac motion estimation is performed using respiratorycompensated images with cardiac gating. The motion estimation algorithm is based on a deformable intensitybased 3D-3D image registration method. Combining the obtained motion vector fields for respiratory and cardiac motion allows us to compensate motion for any arbitrary respiratory and cardiac target phases.

Results: Either 5D double-gated or respiratory-compensated plus cardiac-gated images both contain strong streak artifacts and high noise levels. Our 5D MoCo algorithm is able to significantly improve the image quality while maintaining the same high temporal resolution for respiratory and cardiac motion as achieved with simple double gating. Because all sparse projection streak artifacts are removed, small structures can be delineated even in areas where motion is high. The noise level of patient data is the same as that of 3D CBCT due to making use of 100 % of the projection data for each reconstructed frame.

Cardio-Respiratory Motion Compensation for 5D Thoracic CBCT in IGRT 20 respiratory phases of 10% width, 10 cardiac phases of 20% width



For all images a window level of -250 HU and a window width of 1400 HU was applied.

Conclusion: This work presents a reconstruction method for true 5D imaging in IGRT. Our patient data demonstrate that good image quality is achievable at identical x-ray dose levels and at acquisition times as for today's 3D CBCT. Treatments of regions close to the heart should be able to benefit from our approach.

PO-0935

Correcting diffusion weighted MR images for signal pile-up and distortions near gas pockets

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Purpose or Objective: Diffusion weighted (DW) MRI is used in RT to improve tumor delineation and monitor treatment response. To minimize scan time, echo-planar imaging (EPI) is employed, but variations in the magnetic field (BO) distort

EPI images due to a low pixel bandwidth in the phaseencoding (PE) direction. Geometric distortions can be corrected using a measured B0 map or by combining EPI images obtained with opposite gradients (ref. 1). However, near gas pockets B0 varies strongly. Here signal pile-up can occur, when signals from distinct, possibly non-neighboring, voxel locations are reconstructed into the same voxel. Our objective is to fully correct DW-EPI images using a combination of the above methods.

Material and Methods: On a 3T MRI (Philips Achieva), we acquired EPI images with opposite PE gradients and a dual gradient echo sequence to map BO. Both EPI images are corrected for geometric distortions by the standard correction method using the B0 map. For the new correction method, the B0 map also identifies voxels containing signal pile-up. The distortion-corrected images are averaged into a single image rejecting voxels with signal pile-up. These voxels contain data from only one EPI image. We demonstrated the correction method in a water phantom including an air cavity. The PE gradients had band widths ranging from 6 to 17 Hz/mm, comparable to clinical protocols. The corrected image was compared to raw EPI images and images corrected with the standard method. In a region-of-interest containing only pure water and signal pileup, improvement was quantified as signal homogeneity using the coefficient of variation (CoV defined as standard deviation divided by signal mean). We applied the same method in two patients (prostate and rectal cancer), who underwent an MRI exam before radiotherapy and compared the raw images with the results of the standard correction and our full correction.

Results: With the standard correction method, distortions and intensity variations were removed in the EPI phantom images, but signal pile-up and signal loss were still visible. These were strongly reduced in our method, which was confirmed by the change in CoV in regions with signal pileup. Here, the coefficient was 0.34 and 0.35 for the raw EPI image and B0 corrected image, respectively, and decreased to 0.12 after applying the proposed correction. Patient data are shown in the figure below. Here, rectal gas was present causing distortions and clear signal pile-up in the EPI images. After applying the correction, the signal pile-up was removed resulting in improved images.



Figure: Images of 2 patients with rectum cancer (top row) and with prostate cancer (bottom row). Gas pockets are clearly visible on the high-resolution T2w images and cause variations in 80. Resulting signal pile-up in the EPI images is indicated by arrows. The black regions on the corrected EPI images result from an amplitude threshold on the 80 signal to mask out regions where 80 is unknown.

Conclusion: Our method has shown improvements in correcting EPI images, both in phantom and clinical data. It does not only correct for geometric distortions, but also for possible signal pile-up near gas pockets. Corrected DW-EPI images can improve tumor delineation and response monitoring near these regions.

Ref 1: Jezzard, P., NeuroImage 62 (2012), 648-651

PO-0936

Evolved Grow-cut: A PET based segmentation algorithm for heterogeneous tumors

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