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Appendix

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This file is a(n) Appendix of:

Artificial Intelligence-Driven Single-Shot PET Image Artifact Detection and Disentanglement: Toward Routine Clinical Image Quality Assurance

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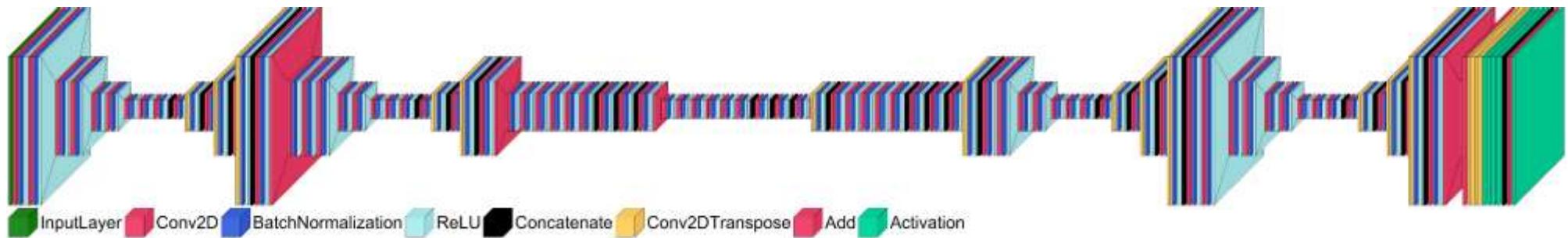
Image preprocessing

Different image preprocessing steps were taken for practical deep neural network training and to reproduce the quantitative PET values¹⁻⁶. First, to reduce the PET intensity dynamic range due to varying injected activity, decay factors, and patient weights, both CT-ASC and non-ASC PET images were converted to standardized uptake values (SUVs)¹⁻⁶. Then, based on 90% of the histogram of all data for further dynamic range reduction, CT-ASC and non-ASC PET images were divided by empiric factors of 5 and 2, respectively¹⁻⁶. Hence, most values were in the range [0-1], and we did not apply any clipping on the values to ensure reproducibility in a clinical context¹⁻⁶.

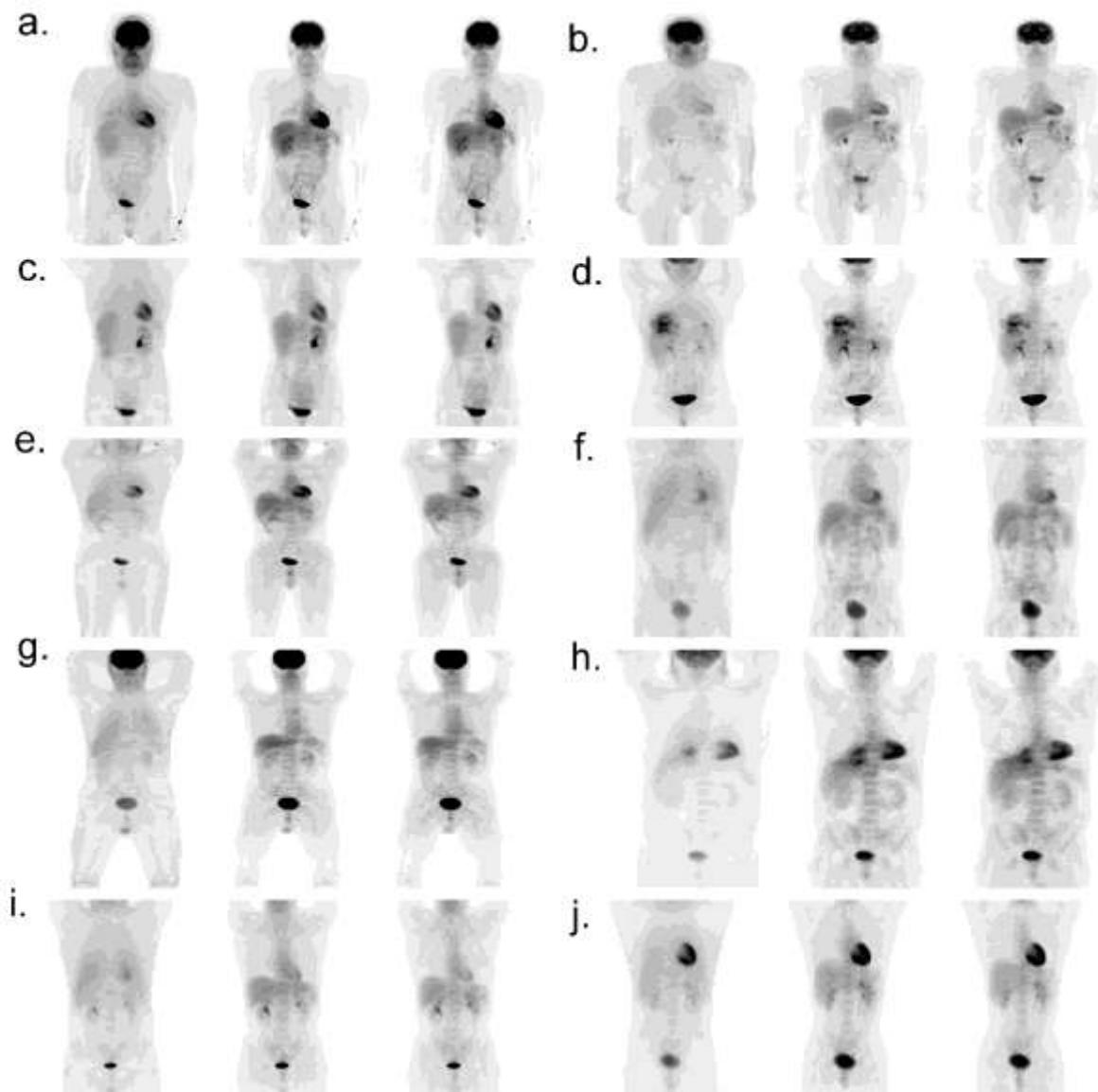
PET-QA-NET Deep Neural Network

We employed the modified U²-Net architecture¹⁻⁷ that utilizes residual U-blocks in a U-shaped architecture¹⁻⁷. In U²-Net, unlike the U-Net-based network, the image resolution is not sacrificed with successive downsampling. Instead, the general U-structure of the U-Net was kept, and each block used another U-Net with a symmetric U-Shaped form known as ReSidual U-block (RSU)¹⁻⁶. This RSU block allows multiscale feature extraction using a mixture of receptive fields. U²-Net⁷ uses deep supervision to include training loss in all scales for further local and global contextual information extraction⁷. The modified U²-Net was trained in a 2D manner with an Adam optimizer, a learning rate of 0.001, an L2-norm loss, and a weight decay of 0.0001¹⁻⁶. During model development and evaluation, we followed various steps adopted from best practice guidelines in nuclear medicine for AI algorithmic development^{8, 9}. Quantitative metrics at the image and region level were calculated on clean test sets for model evaluation¹⁻⁶. Furthermore, a subset of images (100 patients) with artifacts was used for qualitative analysis by physicians.

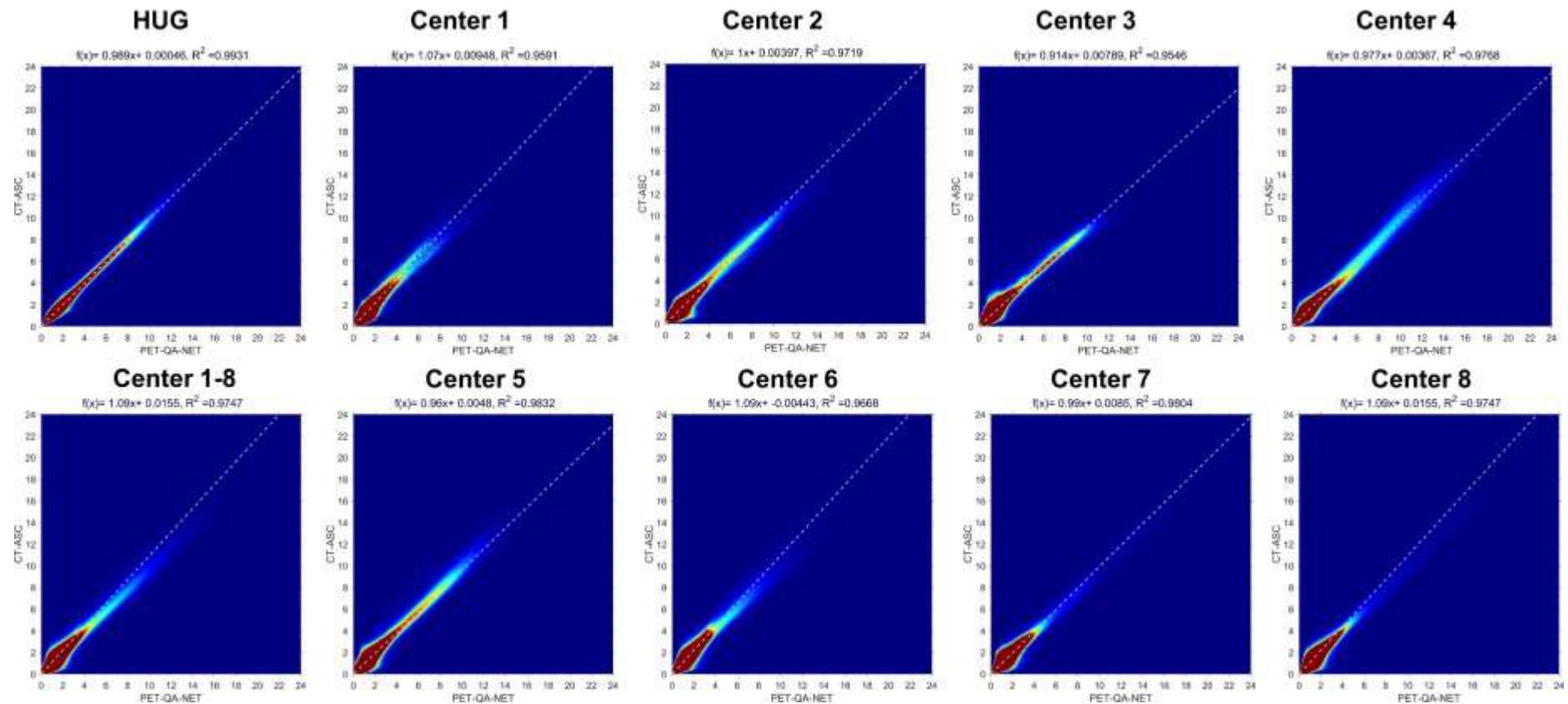
Supplemental Figures



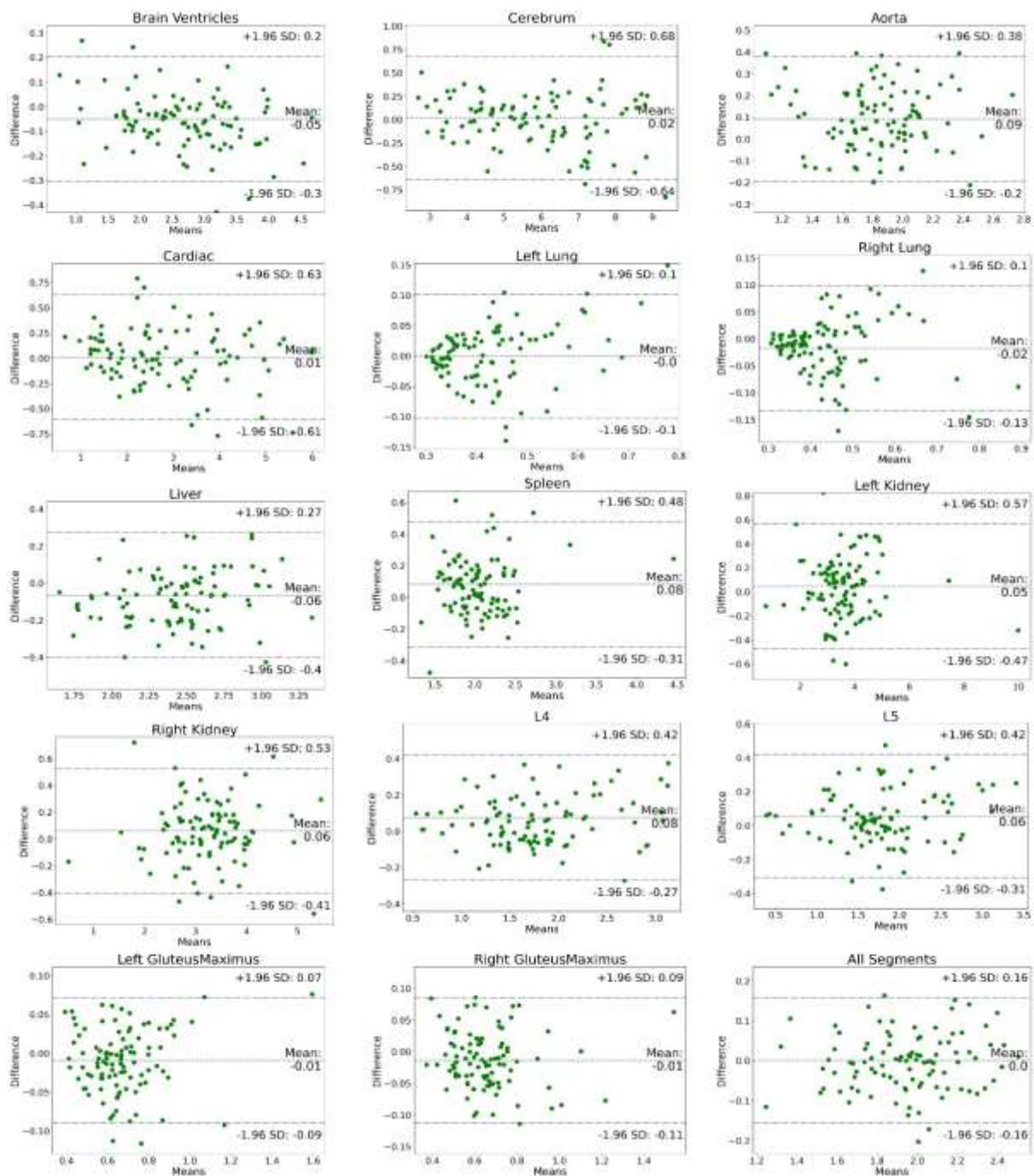
Supplemental Figure 1. U2Net architecture implemented in this study.



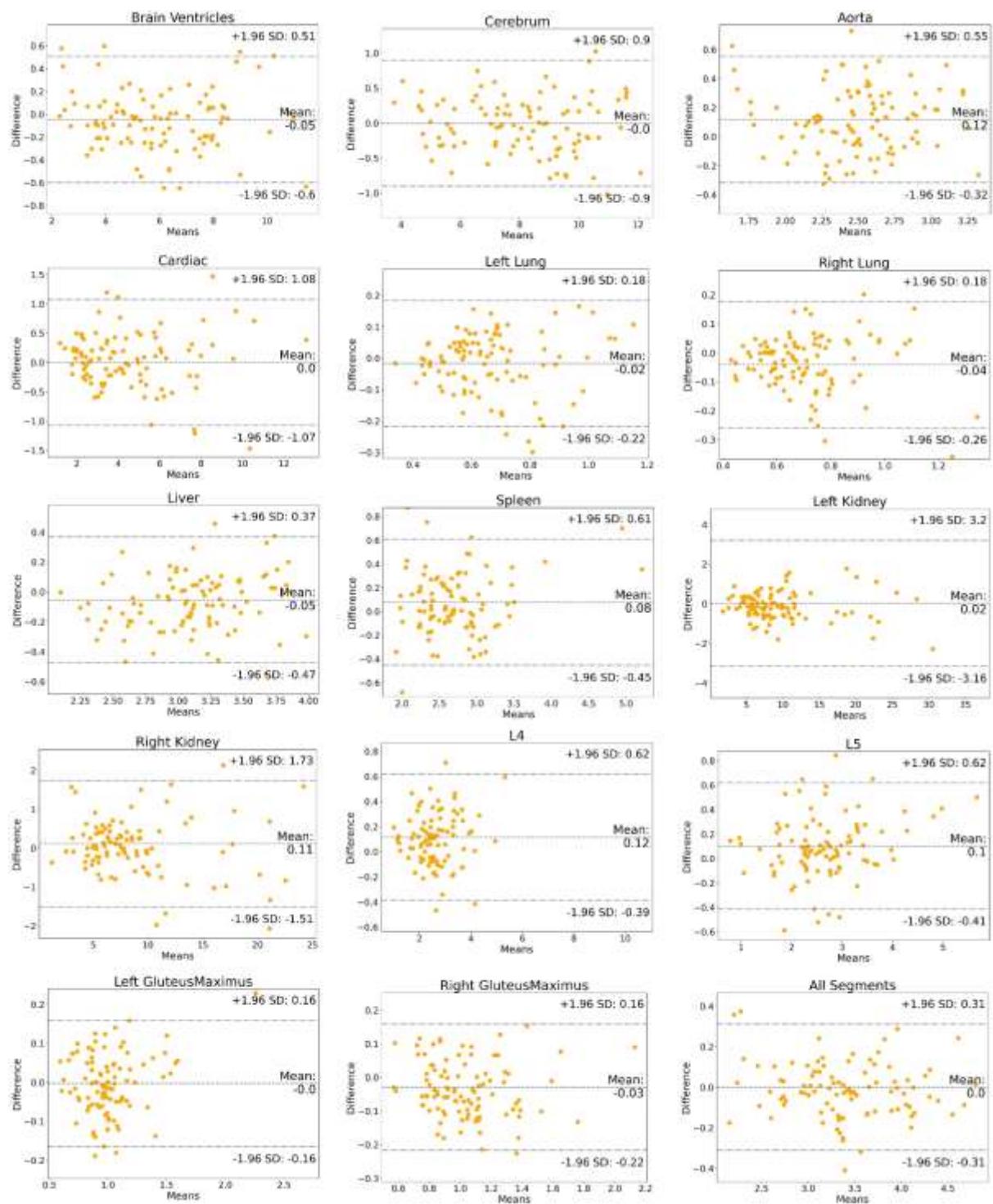
Supplemental Figure 2. MIPs of 8 different centers Non-ASC, CT-ASC, PET-QA-NET from left to right.



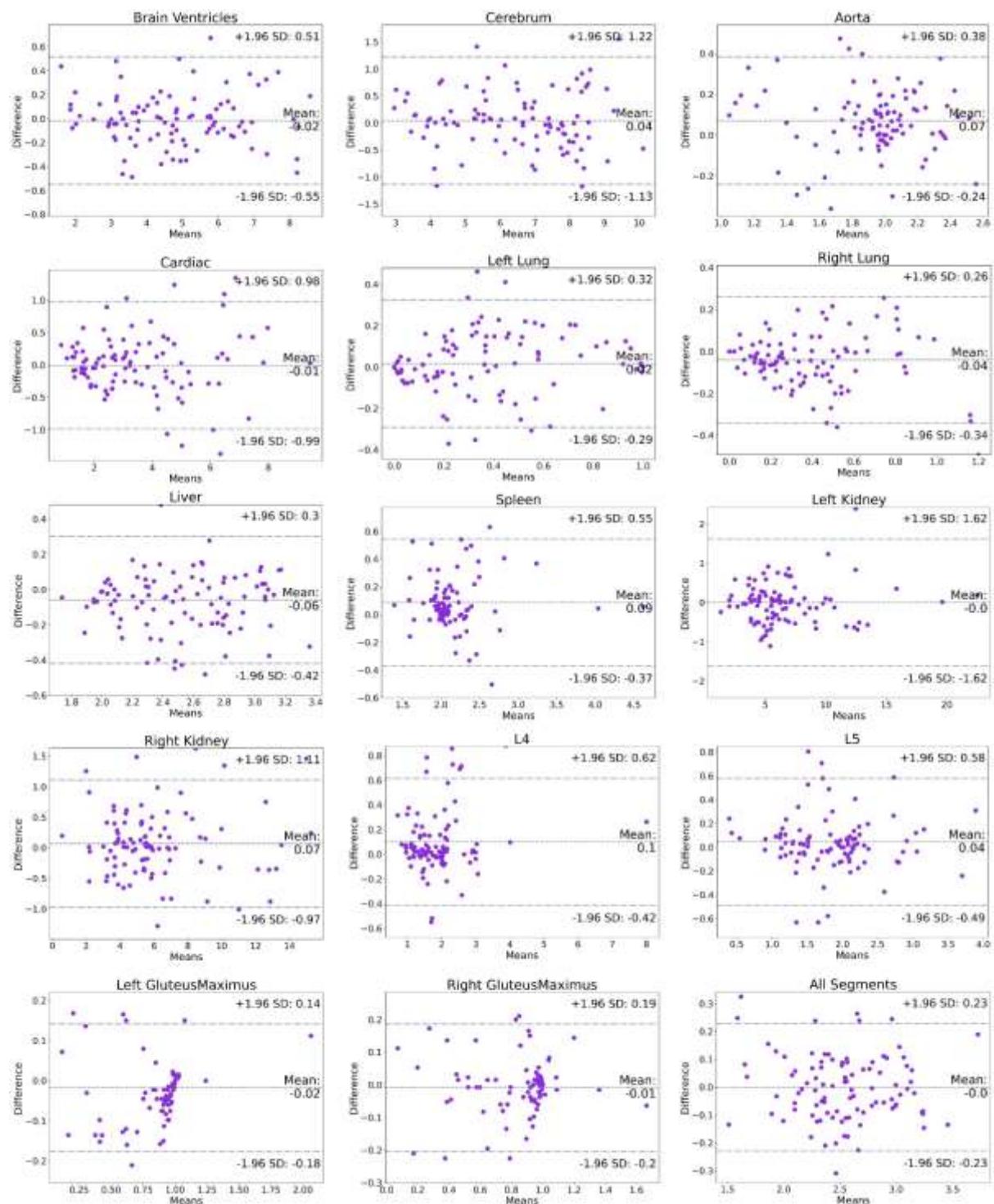
Supplemental Figure 3. Joint histogram of different centers.



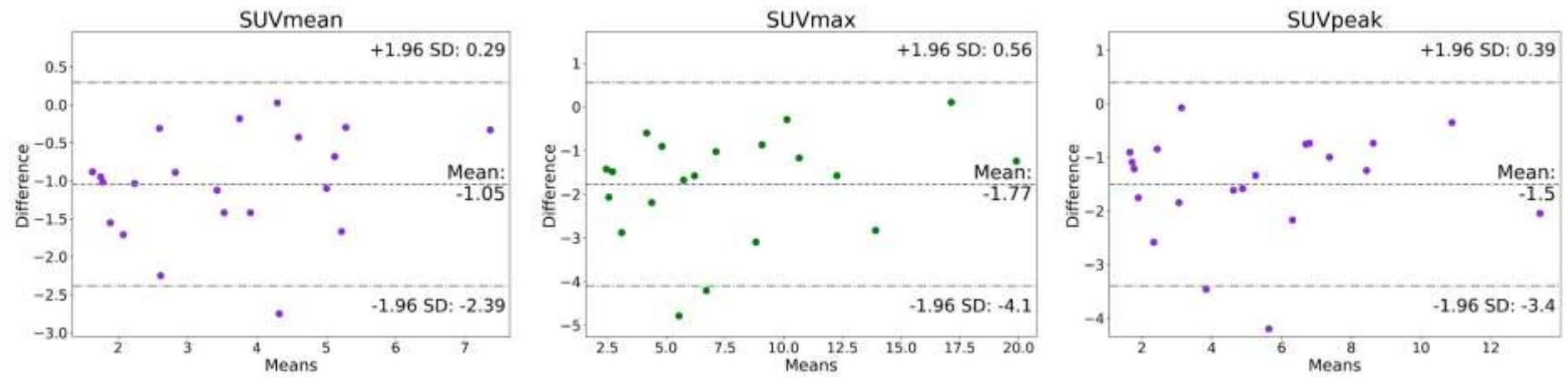
Supplemental Figure 4. Bland Altman of SUV_{mean} for different regions.



Supplemental Figure 5. Bland Altman of SUV_{max} for different regions.



Supplemental Figure 6. Bland Altman of SUV_{peak} for different regions.



Supplemental Figure 7. Bland Altman for SUVs for Motion artifact comparison between CT-ASC and PET-QA-NET.

Supplemental Tables

Supplemental Table 1. Patient demographics and PET/CT image acquisition parameters in the different centers for transfer learning and an external test set ^{1-6,10-22}.

		Centre #1	Centre #2	Centre #3	Centre #4	Centre #5	Centre #6	Centre #7	Centre 8
Demographic	Sex (F/M/NA)	29/44/0	15/31/0	12/39/2	2/23/50	27/37/0	40/57/0	18/121/0	27/120/0
	Age	64.58 ± 12.12	63.83 ± 8.63	61.65 ± 7.25	56.43 ± 7.22	49.9 ± 21.58	67.31 ± 10.58	56.58 ± 9.62	58.88 ± 10.06
Scanner	PET	GE-Discovery ST	GE-Discovery LS	GE-Discovery STE	GE-Discovery STE	Siemens Biograph 6	GE-Discovery 690	GE-Discovery STE	GE-Discovery RX
	Time to Scan	76.04 ± 18.45	82.22 ± 22.69	100.49 ± 30.71	99.44 ± 30.93	96.85 ± 14.04	71.98 ± 18.24	86.86 ± 20.53	82.22 ± 16.31
PET Acquisition and Reconstruction Parameters	Time per Bed	3.08 ± 0.99	3.48 ± 1.35	3.21 ± 0.52	3.21 ± 0.51	3.09 ± 0.35	2.49 ± 0.84	3.03 ± 0.17	2.98 ± 0.23
	Scatter Correction	Model-based, convolution subtraction	Model-based, convolution subtraction	Model-based, convolution subtraction	Model-based, convolution subtraction	Model-based	Model-based, convolution subtraction	Model-based	Model-based
	CT-ASC Reconstruction	OSEM	OSEM	OSEM	OSEM	OSEM	3D FORE+FBP	3D IR	OSEM
	Matrix Size	128×128	128×128	256×256	128×128	168×168	192×192	128×128	128×128
	Slice Thickness	3.27	4.25	3.4	3.27	3	3.27	3.27	3.27

Supplemental Table 2. Summary of radiation dose (mGy) to different organs and effective dose (mSv) for HUG dataset from low-dose CT acquisition for PET attenuation/scatter correction calculated by Impact Dosimetry Tool.

Gender	Male						Female					
	Mean	STD	Q1	Median	Q3	Max	Mean	STD	Q1	Median	Q3	Max
Effective dose (mSv)	2.91	2.39	1.43	2.33	3.72	17.30	3.68	2.79	2.02	3.01	4.52	24.53
Year (from background)	1.21	0.99	0.60	0.97	1.55	7.21	1.53	1.16	0.84	1.25	1.88	10.22
CTDI_W (mGy)	3.53	2.41	2.06	3.25	4.45	18.72	3.69	2.39	2.33	3.37	4.58	25.46
CTDI_{vol} (mGy)	4.41	3.01	2.57	4.06	5.56	23.40	4.61	2.99	2.92	4.21	5.73	31.82
DLP (mGy×cm)	790	540	461	728	996	4193	825	535	522	754	1026	570
SSDE (mGy)	3.42	2.53	1.89	2.89	4.50	17.72	3.57	2.35	2.22	3.06	4.38	18.74
Adrenals (mGy)	3.40	2.81	1.66	2.70	4.36	20.38	3.83	2.93	2.09	3.13	4.71	26.15
Bladder (mGy)	3.77	3.10	1.85	3.01	4.82	22.46	4.24	3.21	2.32	3.46	5.20	28.28
Brain (mGy)	3.77	3.10	1.85	3.01	4.82	22.45	4.18	3.16	2.30	3.42	5.13	27.82
Breast (mGy)	-	-	-	-	-	-	3.42	2.56	1.89	2.81	4.18	22.41
Colon (mGy)	3.10	2.57	1.50	2.45	3.94	18.65	3.69	2.83	2.00	3.00	4.54	25.19
Esophagus (mGy)	3.07	2.56	1.47	2.41	3.90	18.57	3.56	2.75	1.92	2.88	4.42	24.64
Extrathoracic Region (mGy)	4.34	3.56	2.15	3.49	5.54	25.79	4.83	3.63	2.66	3.95	5.89	31.79
Eye (mGy)	3.97	3.23	1.99	3.20	5.11	23.72	4.20	3.11	2.35	3.47	5.17	26.91
Gall Bladder (mGy)	3.50	2.89	1.70	2.77	4.47	20.94	4.05	3.10	2.20	3.30	4.99	27.46
Gonads (mGy)	3.95	3.21	1.97	3.18	5.07	23.52	3.59	2.77	1.93	2.90	4.45	24.86
Heart (mGy)	3.88	3.19	1.90	3.09	4.97	23.14	4.22	3.21	2.31	3.46	5.21	28.39
Kidneys (mGy)	3.67	3.01	1.81	2.94	4.68	21.81	4.20	3.17	2.31	3.44	5.14	27.91
Liver (mGy)	3.34	2.75	1.63	2.65	4.27	19.95	3.92	2.98	2.14	3.20	4.83	26.35
Lower Large Intestine (mGy)	2.94	2.45	1.42	2.32	3.74	17.75	3.52	2.70	1.90	2.85	4.34	24.16
Lung (mGy)	3.72	3.06	1.83	2.97	4.76	22.18	4.27	3.23	2.34	3.49	5.24	28.50
Lymphatic Nodes (mGy)	3.48	2.85	1.72	2.79	4.44	20.64	3.99	3.00	2.20	3.27	4.86	26.24
Muscle (mGy)	3.48	2.85	1.72	2.79	4.44	20.64	3.99	3.00	2.20	3.27	4.86	26.24
Oral Mucosa (mGy)	3.94	3.24	1.93	3.14	5.05	23.53	4.47	3.38	2.45	3.66	5.49	29.68
Pancreas (mGy)	3.12	2.60	1.50	2.45	3.96	18.83	3.66	2.83	1.98	2.97	4.54	25.27
Prostate (mGy)	3.74	3.08	1.84	2.98	4.79	22.33	-	-	-	-	-	-
Red Bone Marrow (mGy)	0.82	0.66	0.41	0.66	1.05	4.90	0.95	0.70	0.53	0.78	1.17	5.92
Salivary Glands (mGy)	4.49	3.65	2.24	3.61	5.75	26.70	4.90	3.65	2.72	4.04	6.01	31.80
Skeleton (mGy)	7.76	6.26	3.92	6.27	10.04	46.76	9.18	6.73	5.17	7.59	11.31	56.94
Skin (mGy)	3.53	2.86	1.77	2.85	4.53	21.08	3.99	2.96	2.23	3.29	4.89	25.60
Small Intestine (mGy)	3.24	2.69	1.57	2.55	4.12	19.48	3.80	2.92	2.06	3.09	4.69	26.00
Spleen (mGy)	3.12	2.58	1.52	2.47	3.99	18.67	3.73	2.84	2.03	3.05	4.59	25.20
Stomach (mGy)	3.41	2.81	1.67	2.71	4.37	20.38	4.02	3.05	2.20	3.29	4.95	26.99
Thymus (mGy)	4.31	3.52	2.14	3.45	5.51	25.53	4.67	3.50	2.58	3.82	5.69	30.55
Thyroid (mGy)	5.73	4.62	2.89	4.63	7.44	34.68	6.19	4.56	3.48	5.11	7.62	39.10
Upper Large Intestine (mGy)	3.25	2.70	1.58	2.57	4.15	19.55	3.86	2.95	2.10	3.14	4.74	26.22
Uterus (mGy)	-	-	-	-	-	-	3.92	3.01	2.12	3.19	4.83	26.77

Supplemental Table 3. Summary statistics of image quality metrics reflecting the quantitative accuracy of the estimated tracer uptake for the clean, untouched test set in the HUG dataset, and 80% of each of the 8 other centers' test sets (transfer learning on 20% of each center for initial fine tuning). MAE: mean absolute error, MSE: mean square error, RE (%): percent relative error, ARE (%): percent absolute relative error, SSIM: Structural Similarity Index, PSNR: Peak signal-to-noise ratio.

		MAE	MSE	RE (%)	ARE (%)	SSIM	PSNR
HUG (n=100)	Mean	0.09±0.02	0.03±0.01	-1.19±3.60	9.91±1.53	0.99±0.00	36.31±1.16
	95% CI	0.09 to 0.10	0.02 to 0.03	-1.90 to -0.47	9.60 to 10.22	0.99 to 0.99	36.08 to 36.54
Centre #1 (n=4)	Mean	0.16±0.02	0.06±0.02	10.19±5.28	18.07±2.54	0.94±0.02	34.33±0.89
	95% CI	0.12 to 0.19	0.03 to 0.08	1.79 to 18.59	14.02 to 22.11	0.90 to 0.98	32.91 to 35.74
Centre #2 (n=8)	Mean	0.16±0.01	0.07±0.02	-3.85±6.96	20.08±1.04	0.92±0.04	33.64±0.90
	95% CI	0.145 to 0.17	0.06 to 0.09	-9.67 to 1.97	19.21 to 20.95	0.89 to 0.95	32.89 to 34.39
Centre #3 (n=11)	Mean	0.16±0.02	0.07±0.02	-1.65±2.77	18.47±1.17	0.81±0.02	34.13±1.08
	95% CI	0.15 to 0.17	0.05 to 0.08	-3.51 to 0.21	17.68 to 19.25	0.80 to 0.82	33.41 to 34.86
Centre #4 (n=15)	Mean	0.14±0.02	0.05±0.01	-2.14±4.98	16.53±1.91	0.94±0.02	34.31±0.86
	95% CI	0.13 to 0.15	0.04 to 0.062	-4.89 to 0.61	15.48 to 17.59	0.93 to 0.95	33.83 to 34.79
Centre #5 (n=31)	Mean	0.15±0.03	0.08±0.09	-2.55±3.63	17.15±2.09	0.90±0.03	33.86±1.54
	95% CI	0.14 to 0.16	0.05 to 0.11	-3.88 to -1.22	16.38 to 17.91	0.89 to 0.92	33.30 to 34.43
Centre #6 (n=37)	Mean	0.16±0.03	0.08±0.05	0.39±5.87	17.82±2.41	0.93±0.03	33.43±1.30
	95% CI	0.14 to 0.16	0.06 to 0.10	-1.56 to 2.35	17.01 to 18.62	0.92 to 0.94	32.99 to 33.87
Centre #7 (n=40)	Mean	0.13±	0.05±0.01	-3.06±6.91	15.91±2.12	0.95±0.01	34.60±1.05
	95% CI	0.12 to 0.13	0.040 to 0.05	-5.27 to -0.85	15.23 to 16.59	0.95 to 0.95	34.26 to 34.93
Centre #8 (n=51)	Mean	0.12±0.02	0.06±0.09	-0.15±4.82	14.87±1.83	0.02	34.84±1.55
	95% CI	0.12 to 0.13	0.032 to 0.08	-1.50 to 1.21	14.36 to 15.38	0.95 to 0.96	34.41 to 35.28
All Centers 1-8 (n=197)	Mean	0.14±0.03	0.06±0.06	-1.19±5.73	16.60±2.43	0.93±0.04	34.23±1.39
	95% CI	0.13 to 0.14	0.054 to 0.07	-1.99 to -0.38	16.26 to 16.94	0.92 to 0.93	34.04 to 34.43

Supplemental Table 4. Mean error and absolute mean error of SUV_{peak}, SUV_{max}, and SUV_{mean} across the different VOIs in a different region for 99 clean test sets. CT-ASC was defined as standard of reference for SUV comparison.

	Organ	Mean±SD	CI95%	Mean±SD	CI95%	Mann Witney P-value
SUV _{mean}	Brain Ventricle	0.05 ± 0.13	0 to 0.02	0.11 ± 0.09	0.1 to 0.09	0.682
	Cerebrum	-0.02 ± 0.34	-0.1 to -0.09	0.24 ± 0.24	0.2 to 0.19	0.96
	Aorta	-0.09 ± 0.15	-0.1 to -0.12	0.14 ± 0.1	0.1 to 0.12	<0.05
	Cardiac	-0.01 ± 0.32	-0.1 to -0.07	0.25 ± 0.21	0.2 to 0.21	0.886
	Right Lung	0.02 ± 0.06	0 to 0.01	0.05 ± 0.05	0 to 0.04	0.124
	Left Lung	0 ± 0.05	0 to -0.01	0.05 ± 0.03	0 to 0.04	0.708
	Liver	0.06 ± 0.17	0 to 0.03	0.16 ± 0.11	0.1 to 0.14	0.206
	Spleen	-0.08 ± 0.2	-0.1 to -0.12	0.16 ± 0.15	0.1 to 0.13	0.144
	Left Kidney	-0.05 ± 0.26	-0.1 to -0.1	0.21 ± 0.17	0.2 to 0.17	0.827
	Right Kidney	-0.06 ± 0.24	-0.1 to -0.11	0.18 ± 0.17	0.1 to 0.15	0.584
	L4	-0.08 ± 0.18	-0.1 to -0.11	0.13 ± 0.14	0.1 to 0.1	0.452
	L5	-0.06 ± 0.19	-0.1 to -0.09	0.14 ± 0.14	0.1 to 0.11	0.562
	Right Gluteus Maximus	0.01 ± 0.05	0 to 0	0.04 ± 0.03	0 to 0.03	0.466
	Left Gluteus Maximus	0.01 ± 0.04	0 to 0	0.03 ± 0.03	0 to 0.03	0.524
	All Segments	0 ± 0.08	0 to -0.02	0.13 ± 0.04	0.1 to 0.12	0.854
	Brain Ventricle	0.05 ± 0.28	0 to -0.01	0.24 ± 0.18	0.2 to 0.2	0.835
	Cerebrum	0 ± 0.46	-0.1 to -0.09	0.35 ± 0.3	0.3 to 0.29	0.956
	Aorta	-0.12 ± 0.22	-0.2 to -0.16	0.2 ± 0.15	0.2 to 0.17	0.046
SUV _{max}	Cardiac	0 ± 0.55	-0.1 to -0.11	0.43 ± 0.37	0.4 to 0.36	0.921
	Right Lung	0.04 ± 0.11	0 to 0.02	0.11 ± 0.09	0.1 to 0.09	0.13
	Left Lung	0.02 ± 0.1	0 to 0	0.1 ± 0.06	0.1 to 0.09	0.712
	Liver	0.05 ± 0.22	0 to 0.01	0.2 ± 0.13	0.2 to 0.18	0.436
	Spleen	-0.08 ± 0.27	-0.1 to -0.13	0.22 ± 0.19	0.2 to 0.18	0.312
	Left Kidney	-0.02 ± 1.62	-0.3 to -0.34	0.78 ± 1.43	0.5 to 0.5	0.829
	Right Kidney	-0.11 ± 0.83	-0.3 to -0.28	0.58 ± 0.6	0.5 to 0.46	0.891
	L4	-0.12 ± 0.26	-0.2 to -0.17	0.2 ± 0.19	0.2 to 0.16	0.348
	L5	-0.1 ± 0.26	-0.2 to -0.15	0.2 ± 0.2	0.2 to 0.16	0.416
	Right Gluteus Maximus	0.03 ± 0.1	0 to 0.01	0.08 ± 0.06	0.1 to 0.07	0.246
	Left Gluteus Maximus	0 ± 0.08	0 to -0.01	0.06 ± 0.05	0.1 to 0.05	0.469
	All Segments	0 ± 0.16	0 to -0.03	0.24 ± 0.1	0.2 to 0.22	0.774
	Brain Ventricle	0.02 ± 0.27	0 to -0.04	0.23 ± 0.21	0.2 to 0.19	0.919
	Cerebrum	-0.04 ± 0.6	-0.2 to -0.16	0.43 ± 0.42	0.4 to 0.35	0.933
	Aorta	-0.07 ± 0.16	-0.1 to -0.1	0.13 ± 0.12	0.1 to 0.11	0.073
SUV _{peak}	Cardiac	0.01 ± 0.5	-0.1 to -0.09	0.39 ± 0.35	0.3 to 0.32	0.968
	Right Lung	0.04 ± 0.15	0 to 0.01	0.15 ± 0.11	0.1 to 0.13	0.286
	Left Lung	-0.02 ± 0.16	0 to -0.05	0.15 ± 0.11	0.1 to 0.13	0.787
	Liver	0.06 ± 0.18	0 to 0.02	0.17 ± 0.13	0.1 to 0.15	0.321
	Spleen	-0.09 ± 0.23	-0.1 to -0.13	0.16 ± 0.19	0.1 to 0.12	0.035
	Left Kidney	0 ± 0.83	-0.2 to -0.16	0.48 ± 0.68	0.3 to 0.34	0.897
	Right Kidney	-0.07 ± 0.53	-0.2 to -0.17	0.39 ± 0.37	0.3 to 0.32	0.883
	L4	-0.1 ± 0.26	-0.2 to -0.15	0.17 ± 0.23	0.1 to 0.12	0.361
	L5	-0.04 ± 0.27	-0.1 to -0.1	0.17 ± 0.21	0.1 to 0.13	0.737
	Right Gluteus Maximus	0.01 ± 0.1	0 to -0.01	0.06 ± 0.08	0 to 0.04	0.656
	Left Gluteus Maximus	0.02 ± 0.08	0 to 0	0.05 ± 0.06	0 to 0.04	0.251
	All Segments	0 ± 0.12	0 to -0.02	0.21 ± 0.06	0.2 to 0.2	0.874

Supplemental Table 5. Intra- and inter-reader correlation coefficient ICC (95% CI) for image quality, artifacts, confidence, and lesion detectability in the different regions.

Region	Image Quality	Artifact	Confidence	Lesion
Intra-reader correlation coefficient ICC				
Brain	0.72 (0.59-0.81)	0.79 (0.69-0.86)	0.69 (0.56-0.79)	0.87 (0.81-0.92)
Chest	0.65 (0.50-0.76)	0.74 (0.62-0.83)	0.72 (0.59-0.81)	0.89 (0.84-0.93)
Chest/Abd interface	0.54 (0.36-0.68)	0.74 (0.62-0.82)	0.57 (0.39-0.70)	0.88 (0.82-0.92)
Abdomen	0.62 (0.46-0.74)	0.72 (0.59-0.81)	0.56 (0.39-0.69)	0.91 (0.86-0.94)
Pelvis	0.68 (0.54-0.78)	0.74 (0.62-0.82)	0.64 (0.50-0.75)	0.96 (0.93-0.97)
Extremities	0.71 (0.58-0.80)	0.75 (0.64-0.83)	0.68 (0.54-0.80)	0.85 (0.77-0.90)
All regions	0.67 (0.61-0.72)	0.78 (0.74-0.81)	0.65 (0.60-0.70)	0.90 (0.89-0.92)
Inter-reader correlation coefficient using ICC				
Brain	0.74 (0.66-0.80)	0.80 (0.68-0.90)	0.71 (0.62-0.77)	0.80 (0.74-0.85)
Chest	0.58 (0.46-0.67)	0.71 (0.58-0.78)	0.53 (0.40-0.64)	0.56 (0.29-0.66)
Chest/Abd interface	0.58 (0.50-0.65)	0.54 (0.45-0.65)	0.52 (0.44-0.61)	0.58 (0.42-0.72)
Abdomen	0.61 (0.47-0.73)	0.56 (0.48-0.67)	0.51 (0.40-0.63)	0.67 (0.55-0.76)
Pelvis	0.53 (0.39-0.63)	0.60 (0.45-0.74)	0.44 (0.35-0.55)	0.66 (0.56-0.77)
Extremities	0.67 (0.58-0.74)	0.76 (0.59-0.89)	0.53 (0.39-0.63)	0.28 (0.10-0.45)
All regions	0.66 (0.60-0.69)	0.68 (0.62-0.74)	0.53 (0.48-0.59)	0.61 (0.51-0.72)

Supplemental Table 6. Comparison of image quality between CT-ASC and PET-QA-NET generated PET images.

Region	Image Quality	p-value*	p-value [#]	p-value%
Head and Neck	Excellent	0.552		
	High	0.636		
	Average	0.633	0.340	0.313
	Poor	<0.05		
Chest	Very poor	0.406		
	Excellent	0.130		
	High	0.760		
	Average	0.527	0.660	0.623
Chest/Abd Int	Poor	0.992		
	Very poor	0.990		
	Excellent	0.096		
	High	<0.001		
Abdomen	Average	<0.05	<0.005	<0.002
	Poor	<0.05		
	Very poor	0.881		
	Excellent	0.732		
Pelvis	High	0.342		
	Average	0.761	0.854	0.543
	Poor	0.417		
	Very poor	0.999		
Extremities	Excellent	0.311		
	High	0.359		
	Average	0.993	0.591	0.487
	Poor	0.231		
All regions	Very poor	0.999		
	Excellent	0.549		
	High	0.223		
	Average	0.438	0.479	0.325
	Poor	0.147		
	Very poor	0.999		
	Excellent	0.430		
	High	0.154		
	Average	0.387	0.442	0.411
	Poor	<0.05		
	Very poor	0.652		

P-value* is based on the McNemar test; P-value[#] is based on the marginal homogeneity test; P-value% is based on a generalized linear model after adjustment of ICC for all raters (CT-ASC and PET-QA-NET) and readers (1 and 2);

Supplemental Table 7. Comparison of confidence between CT-ASC and PET-QA-NET.

Region	Confidence	P-value*	P-value [#]	P-value [%]
Head and Neck	Excellent	0.834	0.799	0.704
	High	0.448		
	Average	0.485		
	Poor	0.999		
	Very poor	0.311		
Chest	Excellent	0.321	0.835	0.698
	High	0.877		
	Average	0.433		
	Poor	0.699		
	Very poor	0.990		
Chest/Abd Int	Excellent	0.406	0.076	<0.05
	High	<0.01		
	Average	<0.01		
	Poor	0.515		
	Very poor	0.990		
Abdomen	Excellent	0.773	0.625	0.487
	High	0.165		
	Average	0.133		
	Poor	0.699		
	Very poor	0.991		
Pelvis	Excellent	0.999	0.148	<0.05
	High	<0.02		
	Average	<0.01		
	Poor	0.992		
	Very poor	0.881		
Extremities	Excellent	0.502	0.085	<0.05
	High	<0.05		
	Average	<0.05		
	Poor	0.649		
	Very poor	0.560		
All regions	Excellent	0.861	<0.05	<0.05
	High	<0.05		
	Average	<0.02		
	Poor	0.777		
	Very poor	0.314		

P-value* is based on the McNemar test; P-value[#] is based on the marginal homogeneity test; P-value[%] is based on a generalized linear model after adjustment of ICC for all raters (CT-ASC and PET-QA-NET) and readers (reader 1 and 2);

Supplemental Table 8. Comparison of the occurrence of artifacts between CT-ASC and PET-QA-NET.

Region	Artifact	P-value*	P-value [#]	P-value%
Head and Neck	None	0.194	0.741	0.664
	Minor	0.395		
	Moderate	0.999		
	Major	0.311		
	Unacceptable	0.990		
Chest	None	<0.02	0.190	0.098
	Minor	0.066		
	Moderate	0.173		
	Major	0.560		
	Unacceptable	0.990		
Chest/Abd Int	None	<0.001	<0.001	<0.001
	Minor	0.093		
	Moderate	<0.001		
	Major	<0.005		
	Unacceptable	0.998		
Abdomen	None	0.876	0.938	0.689
	Minor	0.636		
	Moderate	0.406		
	Major	0.881		
	Unacceptable	0.998		
Pelvis	None	0.272	0.818	0.754
	Minor	0.172		
	Moderate	0.787		
	Major	0.891		
	Unacceptable	0.787		
Extremities	None	<0.05	0.228	0.106
	Minor	0.552		
	Moderate	<0.05		
	Major	0.649		
	Unacceptable	0.560		
All regions	None	0.084	0.071	<0.05
	Minor	0.692		
	Moderate	<0.02		
	Major	0.306		
	Unacceptable	0.562		

P-value* is based on the McNemar test; P-value[#] is based on the marginal homogeneity test; P-value% is based on a generalized linear model after adjustment of ICC for all raters (CT-ASC and PET-QA-NET) and readers (reader 1 and 2);

Supplemental Table 9. Comparison of the number of detected lesions between CT-ASC and PET-QA-NET.

Region	Number	P-value [#]	P-value [%]
Brain	0		
	1		
	2		
	3	0.060	<0.05
	4		
Chest	5≤		
	0		
	1		
	2		
	3	0.337	0.219
Chest/Abd Int	4		
	5≤		
	0		
	1		
	2		
Abdomen	3	0.401	0.378
	4		
	5≤		
	0		
	1		
Pelvis	2		
	3	0.412	0.391
	4		
	5≤		
	0		
Extremities	1		
	2		
	3	0.093	0.069
	4		
	5≤		
All regions	0		
	1		
	2		
	3	0.523	0.398
	4		
	5≤		
	0		
	1		
	2		
	3	0.972	0.891
	4		
	5≤		

P-value[#] is based on the linear-by-linear association using the Mantel-Haenszel test; P-value[%] is based on a generalized linear model after adjustment of ICC for all raters (CT-ASC and PET-QA-NET) and readers (reader 1 and 2).

Supplemental Table 10. Mean error and absolute mean error of SUV_{peak} , SUV_{max} , and SUV_{mean} across the different VOIs for different lesions in the chest/abdomen regions (motion artifact cases). CT-ASC were defined as reference for SUVs comparison.

SUV	Mean	95% CI	Mean	95% CI	P-value
SUV_{max}	-1.77±1.22	-2.33 to -1.22	1.78±1.22	1.24 to 2.34	0.0001
SUV_{mean}	-1.05±0.70	-1.37 to -0.73	1.05±0.70	0.73 to 1.37	0.0001
SUV_{peak}	-1.50±1.00	-2.00 to -1.05	1.50±1.00	1.10 to 1.11	0.0001

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