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Advancing learned algorithms for 2D X-ray computed tomography

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Advancing learned algorithms for 2D X-ray computed tomography

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Contents

1	Introduction	1
1.1	Computed tomography	1
1.1.1	General introduction	1
1.1.2	Tomographic reconstruction as an ill-posed inverse problem	3
1.2	Challenges for computed tomography	5
1.2.1	Objects	5
1.2.2	Imaging	5
1.3	Machine learning	12
1.4	Machine learning for computed tomography	15
1.5	Research questions	18
2	Tailoring CT Acquisitions in the FleX-ray Lab	25
2.1	Functionalities of the FleX-ray scanner	25
2.2	Extensions to the FleX-ray scanner	26
2.2.1	Sample stage disks	26
2.2.2	Sample stage cylinders	27
2.2.3	Beam filtration	30
2.3	Guideline for choosing suitable acquisition parameters	33
3	Beam filtration for object-tailored X-ray CT of multi-material cultural heritage objects	37
3.1	Background	40
3.1.1	Polychromatic X-ray spectrum and metal artifacts	40
3.1.2	Contrast between materials	41
3.1.3	Noise and detector saturation	41
3.2	Methods	42
3.2.1	Concept of beam filtration	42
3.2.2	Hard- and software	43
3.3	Experiments and results	45
3.3.1	The case studies	46
3.3.2	General concept of steps for an object-tailored CT scan	56
3.4	Discussion and conclusion	57

4	2DeteCT - A large 2D expandable, trainable, experimental Computed Tomography dataset for machine learning	61
4.1	Methods	64
4.1.1	X-ray computed tomography scanner	64
4.1.2	Experimental design	64
4.1.3	Data acquisition	71
4.1.4	Computational processing	72
4.2	Data records	76
4.3	Technical validation	77
4.4	Usage notes	77
4.4.1	Raw projection data	77
4.4.2	Reconstructions and segmentations	77
4.4.3	Expansion possibilities	77
4.4.4	Further usage	78
4.5	Code availability	78
5	Learned denoising with simulated and experimental low-dose CT data	81
5.1	Methods	83
5.1.1	Related work	83
5.1.2	Noise simulation	84
5.1.3	Training data	86
5.1.4	Method development	89
5.1.5	Comparison metrics	89
5.1.6	Computational experiments	90
5.2	Results and discussion	91
5.2.1	Empirical selection of noise level	91
5.2.2	Sinogram denoising	92
5.2.3	Optimization in the reconstruction domain: mapping directly from Sinogram to Reconstruction	96
5.3	Discussion and conclusions	96
5.4	Code availability	99
6	Benchmarking Learned Algorithms for Computed Tomography Image Reconstruction Tasks	101
6.1	Related work	103
6.2	Data-driven CT reconstruction	104
6.2.1	Tomographic reconstruction as a linear inverse problem	104
6.2.2	Data-driven methods for tomographic reconstruction	105
6.3	Benchmark design	107
6.3.1	CT Image reconstruction tasks	109
6.3.2	Pipeline	110
6.3.3	Performance metrics	111
6.4	Numerical experiments	112
6.4.1	Pre-processing	112
6.4.2	Evaluated methods	112

6.4.3	Training details	115
6.5	Results and discussion	117
6.5.1	Relevance and difficulty of CT image reconstruction tasks . . .	117
6.5.2	Quantitative and qualitative analysis of the evaluated methods	123
6.5.3	Limitations and broader impact	124
6.5.4	Code and data availability	126
6.6	Conclusion	126
7	Conclusion and outlook	129
7.1	Contributions and limitations	129
7.2	Outlook	132
	Bibliography	135
	List of publications	155
	Samenvatting	157
	Curriculum Vitae	165
	Acknowledgements	167