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## Multi-dimensional feature and data mining

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# **Multi-dimensional feature and data mining**

Theodoros Georgiou

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# Multi-dimensional Feature and Data Mining

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# Contents

<b>Acknowledgements</b>	<b>i</b>
<b>1 Introduction</b>	<b>1</b>
1.1 Background . . . . .	1
1.1.1 Feature extraction of CFD simulation output . . . . .	2
1.1.2 Computer vision . . . . .	3
1.2 Research questions . . . . .	4
1.3 Dissertation outline . . . . .	6
1.4 Dissertation contributions . . . . .	7
1.5 Other work by the author . . . . .	7
<b>2 Deep learning and traditional approaches for high dimensional data</b>	<b>9</b>
2.1 Introduction . . . . .	10
2.2 Deep learning . . . . .	11
2.2.1 Basic deep learning methods . . . . .	12
2.2.2 Deep learning for high dimensional data . . . . .	20
2.3 Traditional methods . . . . .	23
2.3.1 Object surface features . . . . .	24
2.3.2 Volume features . . . . .	29
2.3.3 Spatio-temporal features . . . . .	31
2.4 Datasets and benchmarks . . . . .	38
2.4.1 Object understanding . . . . .	39
2.4.2 Scene understanding . . . . .	40



2.4.3	Video understanding . . . . .	43
2.4.4	Other datasets . . . . .	46
2.5	Research areas . . . . .	47
2.5.1	Object classification and recognition . . . . .	47
2.5.2	Semantic segmentation . . . . .	50
2.5.3	Object detection . . . . .	53
2.5.4	Human action classification . . . . .	57
2.5.5	Other areas . . . . .	62
2.6	Discussion . . . . .	64
2.6.1	Major challenges . . . . .	65
2.6.2	Future work . . . . .	66
2.7	Conclusions . . . . .	66
<b>3</b>	<b>Deep learning for computational fluid dynamics simulation output</b>	<b>69</b>
3.1	Introduction . . . . .	70
3.1.1	Computational fluid dynamics simulations . . . . .	70
3.1.2	Convolutional neural networks . . . . .	71
3.2	Related work . . . . .	72
3.2.1	Flow field pattern recognition . . . . .	72
3.2.2	Convolutional neural networks for CFD simulation output . . . . .	73
3.3	Dataset collection . . . . .	73
3.3.1	Example creation . . . . .	73
3.3.2	Training tasks . . . . .	75
3.4	Network architecture and training details . . . . .	76
3.4.1	General network architecture . . . . .	76
3.4.2	Prediction networks . . . . .	81
3.4.3	Activation functions . . . . .	81
3.4.4	Training details . . . . .	81
3.4.5	Multi task training . . . . .	82
3.5	Experiments . . . . .	84
3.5.1	Activation functions . . . . .	85
3.5.2	Input handling schemes . . . . .	86
3.5.3	Reconstruction . . . . .	87
3.5.4	Fusion stage . . . . .	88
3.5.5	Comparison to a k-NN regressor . . . . .	88
3.6	Conclusion . . . . .	89

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<b>4</b>	<b>Comparing deep learning and hand crafted features for simulation data</b>	<b>91</b>
4.1	Introduction . . . . .	92
4.2	Related work . . . . .	93
4.3	Hand crafted features . . . . .	94
4.4	Deep learning approaches . . . . .	96
4.4.1	Methodology . . . . .	96
4.4.2	Implementation and training details . . . . .	98
4.5	Dataset . . . . .	98
4.6	Experiments . . . . .	99
4.7	Conclusion . . . . .	105
<b>5</b>	<b>Clifford convolution inspired orientation equivariant CNNs</b>	<b>109</b>
5.1	Introduction . . . . .	110
5.2	Related work . . . . .	111
5.3	Clifford convolutions and calculation of rotation angles . . . . .	112
5.4	Layer construction . . . . .	113
5.4.1	Forward pass computations . . . . .	113
5.4.2	Back propagation . . . . .	115
5.5	Experiments . . . . .	117
5.5.1	Datasets and ground truth . . . . .	117
5.5.2	Networks . . . . .	118
5.5.3	MNIST-rot . . . . .	119
5.5.4	Enriched MNIST-rot . . . . .	121
5.5.5	Vehicle Orientation . . . . .	124
5.5.6	Computational complexity . . . . .	126
5.6	Conclusion and future work . . . . .	127
<b>6</b>	<b>Norm Loss: Regularizing artificial neural networks</b>	<b>129</b>
6.1	Introduction . . . . .	130
6.2	Related work . . . . .	131
6.3	Preliminaries . . . . .	133
6.4	Proposed method . . . . .	134
6.4.1	Connection to weight decay . . . . .	135
6.4.2	Computational cost . . . . .	136
6.5	Experiments . . . . .	136
6.5.1	Regularization factor . . . . .	137
6.5.2	Batch size . . . . .	138
6.5.3	CIFAR-10 . . . . .	139
6.5.4	CIFAR-100 . . . . .	141
6.5.5	ImageNet . . . . .	142

6.6	Conclusions . . . . .	143
<b>7</b>	<b>Conclusions</b>	<b>145</b>
7.1	Conclusions . . . . .	145
7.2	Limitations . . . . .	146
7.3	Future work . . . . .	148
<b>APPENDICIES</b>		<b>149</b>
A	Clifford Convolution gradients calculations . . . . .	149
B	Table of abbreviations . . . . .	154
	<b>Bibliography</b>	<b>161</b>
	<b>Summary</b>	<b>205</b>
	<b>Samenvatting</b>	<b>207</b>
	<b>About the author</b>	<b>209</b>