

**S4F3 Graduate Seminar Applied Probability: MC methods & Machine Learning (Sommersemester 2019/20)**  
**Wednesdays 10 c.t., Online via Zoom/Seminarraum 0.006, Endenicher Allee 60**

	<i>Thema</i>	<i>Name</i>	<i>Datum</i>	<i>Literatur</i>
	<b>A. Introduction</b>			
A1.	Deep learning for Applied Mathematicians		22.4.	Higham & Higham
	<b>B. Foundations</b>			<b>Shalev-Shwartz, Part I</b>
B1.	Bias-Complexity Tradeoff		29.4.	5
B2.	VC Dimension		6.5.	6, Anthony-Bartlett
	<b>C. From Theory to Algorithms</b>			<b>Shalev-Shwartz, Part II</b>
C1.	Linear Predictors		13.5.	9
C2.	Boosting		20.5.	10
C3.	Convex learning problems		3.6.	12
C4.	Regularization and stability		10.6.	13
C5.	Stochastic Gradient Descent		17.6.	14
C7.	Support vector machines, kernel methods		24.6.	15, 16
C8.	Nearest neighbour		1.7.	19
C9.	Neural networks		8.7.	20
	<b>D. Mean-field particle systems and neural networks</b>			
D1.	Breaking the curse of dimension with convex NN		Freitag 26.6.	Bach
D2.	Mean field analysis of NN		Freitag 26.6.	Sirignano, Siliopoulos
D3.	Trainability and accuracy of NN: An IPS approach I		Freitag 3.7.	Rotskoff, vanden Eijnden
D4.	Trainability and accuracy of NN: An IPS approach II		Freitag 3.7.	Rotskoff, vanden Eijnden

[Higham & Higham: Deep learning: An introduction for applied mathematicians](#)

[Shalev-Shwartz, Ben-David: Understanding Machine Learning](#)

[Bubeck: Theory of convex optimization for Machine Learning](#)

[Goodfellow, Bengio, Courville: Deep Learning](#)

[Nielsen: Neural networks and deep learning](#)