reading list

(VC dim chapter) Understanding Machine Learning: From Theory to Algorithms, https://www.cs.huji.ac.il/w~shais/UnderstandingMachineLearning/copy.html

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