Instructor	Daniele Durante	
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Objectives	The course will combine methods, applications, theory and computational aspects in Statistical Machine Learning. More specifically, methods will be motivated and evaluated with a focus on applied problems, but the overarching goal will be on the general methodological framework, including theoretical results and statistical properties. The topics will be presented via a careful discussion of the original papers.	
Syllabus	<ol> <li>An Introduction to Statistical and Machine Learning [L1 – L2]</li> <li>Breiman, L. (2001). Statistical modeling: The two cultures. Statistical Science. 16(3): 199–231.</li> <li>Jordan, M.I. (2019). Artificial intelligence—the revolution hasn't happened yet. Harvard Data Science Review. 1(1) + Candes, E., Duchi, J., Sabatti, C. (2019). Statistics and the oncoming AI revolution. Harvard Data Science Review. 1(1) + Donoho, D. (2019) What's missing from today's machine intelligence juggernaut? Harvard Data Science Review. 1(1)</li> <li>Efron, B. (2020). Prediction, estimation, and attribution. Journal of the American Statistical Association. 115(530): 636–655.</li> <li>Regression Trees, Bagging, Boosting and (Causal) Random Forests [L3 – L4 – L5]</li> </ol>	
	<ul> <li>Breiman, L., Friedman, J. H., Olshen, R. A. and Ston Trees. Chapman &amp; Hall.</li> <li>Breiman, L. (1996). Bagging predictors. Machine Learn</li> <li>Breiman, L. (2001). Random forests. Machine Learn</li> <li>Friedman, J., Hastie, T. and Tibshirani, R. (2000). A of boosting. The Annals of Statistics, 28(2): 337–40</li> <li>Wager, S., and Athey, S. (2018). Estimation and E using random forests. Journal of the American Statistics</li> </ul>	earning. 24(2): 123–140. ning. 45(1): 5–32. Additive logistic regression: A statistical view 7. inference of heterogeneous treatment effects
	<ul> <li>3. Bayesian Trees and Bayesian Additive Regression Trees [L6 – L7]</li> <li>Chipman, H. A., George E. I. and McCulloch R. E. (1998). Bayesian CART model search. Journal of the American Statistical Association. 93(443): 935–948.</li> <li>Chipman, H. A., George E. I. and McCulloch R. E. (2010). BART: Bayesian additive regression trees. The Annals of Applied Statistics. 4(1): 266–298.</li> </ul>	
	<ul> <li>4. Conformal Inference/Prediction [L8 – L9]</li> <li>Angelopoulos, A.N., and Bates, S. (2023). Conformal prediction: A gentle introduction. Foundations and Trends® in Machine Learning, 16(4): 494–591.</li> </ul>	
	<ul> <li>5. Topics in Nonparametric Density Regression [L1</li> <li>Quintana, F. A., Müller, P., Jara, A., and MacEach process and related models. <i>Statistical Science</i>, 37(1)</li> </ul>	hern, S. N. (2022). The dependent Dirichlet
	<ul> <li>6. Topics in Bayesian Unsupervised Learning [L11 – L12]</li> <li>Hoff, P. (2021). Additive and multiplicative effects network models. Statistical Science. 36: 34–50.</li> <li>Legramanti, S., Rigon, T., Durante, D. and Dunson, D.B. (2022). Extended stochastic block models with application to criminal networks. Annals of Applied Statistics. 16(4): 2369–2395.</li> <li>Gopalan, P., Hofman, J.M., and Blei, D.M. (2015). Scalable recommendation with hierarchical Poisson factorization. In UAI 2015 Proceedings, pp. 326-335. + Gopalan, P., Ruiz, F.J., Ranganath, R., and Blei, D. (2014). Bayesian nonparametric Poisson factorization for recommendation systems. In AISTATS 2014 Proceedings, pp. 275-283.</li> </ul>	
	Additional references to specific articles will be suggested during the course.	

GRADING The evaluation is based on an individual project in which the student is asked to provide a critical and thoughtful discussion of an interesting topic considered during the course.