

**PLACE AND TIME:** Room DIA004, Mondays 11:15-13:00 and 14:15-16:00

**INSTRUCTOR:** Ali H. Sayed, Email: [ali.sayed@epfl.ch](mailto:ali.sayed@epfl.ch)

**TEACHING ASSISTANT:** Stefan Vlaski, E-mail: [stefan.vlaski@epfl.ch](mailto:stefan.vlaski@epfl.ch)

**COURSE MATERIAL:** Lecture notes authored and distributed by the instructor for exclusive use by students enrolled in the class.

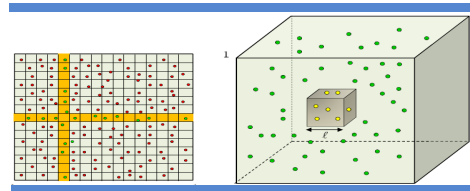
**PRE-REQUISITES:** It is recommended that students have some familiarity with matrix theory, linear algebra, and probability. Supplemental material on these topics is provided by the instructor as needed.

**GRADING:** 4 homework assignments including some computer projects (40%) and two take-home exam assignments worth 30% each.

**EXAMS:**

- Take-home exam I (out on April 1, 2019 after class; return on April 8, 2019 before class).
- Take-home exam II (out on May 27, 2019 after class; return on May 30, 2019 before 4PM).

**TOPICS:** In this course, students learn to master tools, algorithms, and core concepts related to inference from data, data analysis, and adaptation and learning theories. *Emphasis is on the theoretical underpinnings and statistical limits of learning theory.* In particular, the course covers topics related to optimal inference, regularization, proximal techniques, online and batch methods, stochastic learning, generalization and statistical learning theory, Bayes and naive classifiers, nearest-neighbor rules, clustering, self-organizing maps, decision trees, logistic regression, discriminant analysis, Perceptron, support vector machines, kernel methods, bagging, boosting, random forests, cross-validation, and principal component analysis.



| LECTURE | TASK           | DATE           | TENTATIVE TOPIC(S)  |
|---------|----------------|----------------|---|
| 1       |                | Feb. 18        | Vector Differentiation. Convex Functions. Proximal Operator.            |
| 2       |                | Feb. 25        | Deterministic and Stochastic Optimization.                              |
| 3       | HW1 due        | Mar. 4         | Motivation (Inference, Classification, Clustering). Maximum Likelihood. |
| 4       |                | Mar. 11        | Expectation-Maximization. Mixture Models.                               |
| 5       |                | Mar. 18        | Mean-Square-Error Inference. Linear Regression. Least-Squares.          |
| 6       | HW2 due        | Mar. 25        | Regularization. LASSO. Basis Pursuit.                                   |
| 7       | Exam I out     | Apr. 1         | Bayesian Inference. Discriminant Analysis.                              |
| 8       | Exam I due     | Apr. 8         | Logistic Regression. Cross Validation. Perceptron.                      |
| 9       |                | Apr. 15        | Support Vector Machines. Naïve Bayes. Nearest-Neighbor Rule.            |
|         | <b>HOLIDAY</b> | <b>Apr. 22</b> | <b>EASTER HOLIDAY (NO CLASS)</b>  |
| 10      | HW 3 due       | Apr. 29        | k-means Clustering. Decision Trees.                                     |
| 11      |                | May 6          | Generalization Theory. Kernel-Based Learning. PCA                       |
| 12      |                | May 13         | Neural Networks.  |
| 13      | HW4 due        | May 20         | Deep Networks.  |
| 14      | Exam II out    | May 27         | Convolutional Networks.   |
|         | Exam II due    | May 30         | Before 4 PM.  |