## **Linear Regression**

## **Useful Formulas**

• Pseudoinverse of matrix  $\Phi$ :  $\Phi_p = \left(\Phi^T \Phi\right)^{-1} \Phi^T$ 

## Questions

- 1. What is the role of basis functions in linear regression?
- 2. Can an algorithm doing linear regression learn only linear functions of the inputs?
- 3. When can we solve the linear regression problem exactly (with 0 error)? Why is it not a good idea to do so?
- 4. What is the error we want to minimize when doing linear regression?
- 5. What is the least-squares solution? How is it affected by outliers?
- 6. How can we find the least-squares solution when there are too many points to compute the pseudoinverse efficiently?
- 7. What are the bias and the variance for a supervised learning problem?
- 8. What is the link between the error on the validation set increasing with training, and the bias/variance decomposition?
- 9. Given the dataset: <-1, -0.5>, <0,1.1>, <1,3.8>, <2,8.8>, find the least-squares solution for the function:  $y(x, \mathbf{w}) = w_0 + w_1 x$
- 10. 10. Given the dataset: <-1, 0.78>, <0,1>, <1,1.22>, <2,1.52>, find the least-squares solution for  $y(x, \mathbf{w}) = w_0 + w_1 e^{\frac{(x+1)^2}{20}}$
- 11. Given the dataset: <-1, 1.6>, <0,0.95>, <1,1.2>, <2,1.9>, find the least-squares solution for the function:  $y(x, \mathbf{w}) = w_0 + w_1 \frac{1}{1 + e^{-(x+1)}}$