# Exercise 2: Least Squares

## Background reading

The background material for this exercise is Sections 10.1, 13.2 and Appendix II of Ljung (System Identification; Theory for the User, 2nd Ed., Prentice-Hall, 1999).

#### Problem 1:

Consider the least-squares (LS) estimation problem:

$$Y = \Phi\theta + \epsilon$$
.

where  $\Phi$  is the regressor matrix and  $\theta$  is the parameter vector to be estimated

$$Y := \begin{bmatrix} y(0) \\ \vdots \\ y(N-1)) \end{bmatrix}, \qquad \Phi := \begin{bmatrix} \varphi^{\top}(0) \\ \vdots \\ \varphi^{\top}(N-1) \end{bmatrix}, \qquad \theta := \begin{bmatrix} \theta_1 \\ \vdots \\ \theta_p \end{bmatrix}.$$

Assume that the noise,  $\epsilon$ , is zero-mean Gaussian and correlated with  $E\left\{\epsilon\epsilon^{\top}\right\} = R$ . In this exercise we look for a linear estimator  $\hat{\theta}$  of the form,

$$\hat{\theta} = Z^{\top} Y, \tag{2.1}$$

which is unbiased and minimizes its variance. For a given  $\Phi$  show the following:

- 1. For a linear estimator of the form (2.1) to be unbiased we require that  $Z^{\top}\Phi = I$ .
- 2. The covariance matrix of any linear unbiased estimator of the form (2.1) is  $\operatorname{cov}\left\{\hat{\theta}\right\} = Z^{\top}RZ$ .
- 3. The covariance matrix of the best linear unbiased estimator (BLUE)  $\hat{\theta}_Z$  with  $\hat{\theta}_Z = (\Phi^\top R^{-1}\Phi)^{-1}\Phi^\top R^{-1}Y$  is  $\operatorname{cov}\left\{\hat{\theta}_Z\right\} = (\Phi^\top R^{-1}\Phi)^{-1}$ .
- 4. The best linear unbiased estimator  $\hat{\theta}_Z$  exhibits the smallest variance in the class of all unbiased estimators, i.e.  $\operatorname{cov}\left\{\hat{\theta}_Z\right\} \leq \operatorname{cov}\left\{\hat{\theta}\right\}$ .

**Hint:** All covariance matrices are positive semi-definite and in our case we can assume that R is positive definite. The inverse of a positive definite matrix is also positive definite.

### Problem 2:

Show that the following problems can be solved using linear least squares, with the estimate given by

$$\hat{\theta} = (X^T X)^{-1} X^T y. \tag{2.2}$$

For each write down  $\hat{\theta}$ , X, and y.

- 1. fitting the coeficients of a polynomial of degree k,  $p(x) = \sum_{i=0}^{k} a_i x^i$ , to n i.i.d. observations.
- 2. fitting the equation of a circle,  $(x x_c)^2 + (y y_c)^2 = r^2$ , to n i.i.d. observations.

**Hint 1:** Write the expression in matrix form, with the coefficients of the polynomial isolated in a vector.

**Hint 2:** The general equation for a conic section is  $F(x,y) = ax^2 + bxy + cy^2 + dx + ey + f = 0$ , with a, b and c not all zero. A circle is a conic section.

## Problem 3: (MATLAB)

Data was collected on the position of two comets traveling in the solar system. You are provided with the files C-2017-K2.mat and P-2011-NO1.mat, containing xy observations in the orbit plane, in light-years.<sup>1</sup>

The general equation for a conic section is  $F(x,y) = ax^2 + bxy + cy^2 + dx + ey + f = 0$ , with a, b and c not all zero. An ellipse satisfies  $b^2 - 4ac < 0$ , a parabola  $b^2 - 4ac = 0$ , and an hyperbole  $b^2 - 4ac > 0$ .

- 1. Write the linear least squares estimate for the general equation of a conic section.
- 2. Using Matlab estimate the coeficients of the equation for each of the comets.
- 3. Are any of these comets periodic?

**Hint 1:** The equation is satisfied for any scaling of the parameters, you can set f = 1 to avoid a trivial solution.

<sup>&</sup>lt;sup>1</sup>This data was generated from the published orbit elements of the comets, and are not direct observations.