## Part I

## Lagrange multiplier

To look for the optimums of a function :

$$f(t_1, t_2, ..., t_p)$$

of p variables subject to:

$$l(t_1, t_2, ..., t_p) = \text{cte},$$

we compute the partial derivatives of the function :

$$g(t_1, t_2, ..., t_p) = f(t_1, t_2, ..., t_p) - \lambda (l(t_1, t_2, ..., t_p) - \text{cte})$$

with respect to each variable. By canceling the p partial derivatives obtained and considering the single constraint, we obtain a linear system with p+1 equations. The p+1 unknowns are the variables  $(t_1,...,t_p)$  and the Lagrange multiplier  $\lambda$ . The existence of solutions for the linear system is a necessary condition but not sufficient to the existence of an optimum for the function f.

We can generalize the previous approach when the p variables are subject to c constraints. We consider now a new function  $g(t_1, t_2, ...t_p)$  with a linear combination of all constraints whose coefficients are the Lagrange multipliers  $\lambda_1, ..., \lambda_c$  and we solve the linear system with p + c equations associated.

## Part II

## Derivative of a quadratic form with respect to a vector

To look for the principal components, we had to to compute the partial derivatives of both  ${}^t a \Sigma a$  and  ${}^t a a$  with respect to the coefficients  $a_1, ..., a_p$  of the vector a. We denote  $\frac{\partial g(a)}{\partial a} \in \mathbb{R}^p$  the vector whose components are the partial derivatives of g(a) with respect to each component of the vector a:

$$\frac{\partial g\left(a\right)}{\partial a} = \begin{bmatrix} \frac{\partial g\left(a\right)}{\partial a_{1}} \\ \vdots \\ \frac{\partial g\left(a\right)}{\partial a_{j}} \\ \vdots \\ \vdots \\ \frac{\partial g\left(a\right)}{\partial a_{p}} \end{bmatrix}$$

One can show that :

$$\frac{\partial \left( ^{t}a\Sigma a\right) }{\partial a}=2\Sigma a.$$

Indeed:

$$\frac{\partial \begin{pmatrix} {}^t a \Sigma a \end{pmatrix}}{\partial a} = \begin{bmatrix} \frac{\partial^t a}{\partial a_1} \Sigma a \\ \vdots \\ \vdots \\ \frac{\partial^t a}{\partial a_j} \Sigma a \\ \vdots \\ \vdots \\ \frac{\partial^t a}{\partial a_p} \Sigma a \end{bmatrix} + \begin{bmatrix} {}^t a \Sigma \frac{\partial a}{\partial a_1} \\ \vdots \\ {}^t a \Sigma \frac{\partial a}{\partial a_j} \\ \vdots \\ \vdots \\ {}^t a \Sigma \frac{\partial a}{\partial a_p} \end{bmatrix}.$$

One can notice that on each line, both elements are equal since they are of dimension  $1\times 1$  and are the transpose of each other. Thus :

$$\frac{\partial \left( ^{t}a\Sigma a\right) }{\partial a}=2\frac{\partial ^{t}a}{\partial a}\Sigma a,$$

and the derivative of  ${}^ta$  with respect to a is :

with respect to 
$$a$$
 is: 
$$\frac{\partial^{t} a}{\partial a} = \begin{bmatrix} \frac{\partial^{t} a}{\partial a_{1}} \\ \vdots \\ \frac{\partial^{t} a}{\partial a_{j}} \\ \vdots \\ \frac{\partial^{t} a}{\partial a_{p}} \end{bmatrix} = I_{p}$$
 of dimension  $p$ . Similarly we

which the identity matrix of dimension p. Similarly we can show that :  $\frac{\partial^{t} aa}{\partial a} = 2a$