## Mapping functions to n-sphere

Anjishnu Bandyopadhyay January 19, 2016 Let  $f_i(x|\alpha_i)$  i = 0 i = 0 be n + 1 functions we want to map on the surface of an n-sphere. The functions can be completely described by their Fisher distance matrix

$$\Delta_{ij} = \arccos(\int \sqrt{f(x|\alpha_i)f(x|\alpha_j)})$$

We suppose that  $f(x|\alpha_0)$  is the nominal distribution and  $f(x|\alpha_k)$  k=1(k)n are the n variational distributions. At first, we embed  $f(x|\alpha_0)$  onto the south pole of the n-sphere. The position vector of the south pole of the n-sphere is  $I_0 = (0, 0, \dots, x_{00})$  where  $x_{00} = -1$ .

Consequently, we embed the functions  $f(x|\alpha_k)$  k = 1(k)n recursively. We use n-dimensional vectors to denote the position of the functions mapped onto the sphere. The form of the vectors used is shown below

$$X_{1} = (0, 0, 0, \dots, x_{01}, x_{11})$$

$$X_{2} = (0, 0, 0, \dots, x_{02}, x_{12}, x_{22})$$

$$\vdots \dots \vdots$$

$$\vdots \dots \vdots$$

$$X_{k} = (0, 0, 0, \dots, x_{0k}, x_{1k}, x_{2k}, x_{3k}, \dots, x_{k-1,k}, x_{kk})$$

$$\vdots \dots \vdots$$

$$\vdots \dots \vdots$$

$$\vdots \dots \vdots$$

$$X_{n} = (x_{0n}, x_{1n}, \dots, x_{n-1,n}, x_{nn})$$

$$(1)$$

So for  $X_k$ , the first n-k coordinates are 0 and rest of them are embedded on  $\mathbb{R}^k$ . Since these vectors lie on the surface of the *n*-sphere they should satisfy

$$\sum_{i=0}^{k} x_{ik}^2 = 1 \ \forall \ k \tag{2}$$

In addition to this, the position vectors must be defined in such a way that the distance between any two points  $X_k$  and  $X_m$   $k \neq m$  is their Fisher information distance i.e.  $\Delta_{km}$ . Therefore we have

$$x_{ok}^{2} + x_{1k}^{2} + \dots + x_{m+1,k}^{2} + (x_{k-(m+1),k} - x_{0m})^{2} + (x_{k-m,k} - x_{1m})^{2} + (x_{k-(m-1),k} - x_{2m})^{2} + \dots + (x_{k-1,k} - x_{m-1,m})^{2} + (x_{kk} - x_{mm})^{2} = \Delta_{km}^{2}$$
(3)

Since the geometry is spherical we can use Euclidean distances as a measure of the distance between the different points. Combining equations (2) and (3) we get

$$x_{k-(m+1),k} \cdot x_{0m} + x_{k-m,k} \cdot x_{1m} + \dots + x_{k-1,k} \cdot x_{m-1,m} + x_{kk} \cdot x_{mm} = 1 - \frac{\Delta_{km}^2}{2}$$
 (4)

When we embed the points recursively using (4), for  $X_k$ , k equations of the form (4) needs to be solved. These k equations can be written as a matrix equation as shown in (5)

$$\begin{pmatrix} x_{0,k-1} & x_{1,k-1} & \cdots & \cdots & x_{k-1,k-1} \\ 0 & x_{0,k-2} & x_{1,k-2} & \cdots & x_{k-2,k-2} \\ 0 & 0 & \cdots & \cdots & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & x_{01} & x_{11} \\ 0 & 0 & \cdots & 0 & x_{00} \end{pmatrix} \cdot \begin{pmatrix} x_{1k} \\ x_{2k} \\ \vdots \\ x_{k-1,k} \\ x_{kk} \end{pmatrix} = \begin{pmatrix} 1 - \frac{\Delta_{1k}^2}{2} \\ 1 - \frac{\Delta_{2k}^2}{2} \\ \vdots \\ 1 - \frac{\Delta_{k-1,k}^2}{2} \\ 1 - \frac{\Delta_{k}^2}{2} \end{pmatrix}$$
 (5)

Comparing (1) and (5) we notice that the rows of the first matrix are the position vectors  $X_{k-1}, X_{k-2}, \dots, X_1, X_0$  respectively which have been computed using the previous similar k-1 matrix equations. These k equations preserve the distances of  $X_k$  with  $X_0 \cdots X_{k-1}$ . The other (n-k) Fisher distances are preserved by solving the consequent (n-k) matrix equations for  $X_{k+1}...X_n$ . As it can be seen, (5) does not compute  $x_{0k}$ . After solving (5), we compute  $x_{0k}$  by using (2).