regression, such [9]. In this case, we can define the correlation matrix

$$C = \left[ \begin{array}{cc} 1 & \rho \\ \rho & 1 \end{array} \right],$$

and we want to estimate  $\rho$  along with the variances of  $u_i$  and  $v_i$ . Here, it is trivial to ensure that C is positive-definite, by requiring  $-1 < \rho < 1$ , but in higher dimensions, this issue requires more careful consideration.

To ensure that C is positive-definite, you can parameterize the problem in terms of the Cholesky factor L, i.e., C = LL', where L is a lower diagonal matrix with positive diagonal elements. There are q(q-1)/2 free parameters (the non-zero elements of L) to be estimated, where q is the dimension of C. Since C is a correlation matrix, we must ensure that its diagonal elements are unity. An example with q=4 is

```
PARAMETER_SECTION
```

```
matrix L(1,4,1,4) // Cholesky factor
init_vector a(1,6) // Free parameters in C
init_bounded_vector B(1,4,0,10) // Standard deviations
```

## PROCEDURE\_SECTION

```
k = 1
int k=0;
L(1,1) = 1.0;
for(i=2;i<=4;i++)
{
    L(i,i) = 1.0;
    for(j=1;j<=i-1;j++)
        L(i,j) = a(k++);
    L(i)(1,i) /= norm(L(i)(1,i)); // Ensures that C(i,i) = 1
}</pre>
```

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