C++ Functions in Maxliklib Library

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Abstract

The functions in the maxliklib repository are described. Arguments and their definitions are specified, and dependencies of functions are stated.

Keywords: Maximization procedures, quadrature procedures, maximum likelihood

The maxliklib repository consists of C++ functions helpful in estimation related to maximum likelihood. The functions should be appropriate for C++14. They rely on the Armadillo library (Sanderson & Curtin, 2016, 2018) at http://arma.sourceforge.net, the StatsLib library at https://www.kthohr.com/statslib.html, and the Boost library at https:/www.boost.org. Unless otherwise noted, for the library members considered, it is assumed that users have verified that function arguments are valid. Namespaces assumed where relevant are *std*, *arma*, and *boost::math*. The following functions are found in the library.

- adaptpwr.cpp
- addsel.cpp
- berresp.cpp
- conjgrad.cpp
- conjgradn.cpp
- contresp.cpp
- cumresp.cpp

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- rebound.cpp
- rescale.cpp
- trisym.cpp

- truncresp.cpp
- unpack.cpp
- vecsel.cpp
- wmc.cpp

Distributions of Sums of Independent Multinomial Variables

The functions in this section implement a modified and generalized version of the Lord-Wingersky algorithm (Lord & Wingersky, 1984; Thissen et al., 1995). The numerical procedures and their rationale are discussed in lw.pdf.

lw.cpp

The function lw.cpp finds the probability mass function of the sum S of mutually independent Bernoulli random variables X_j, $0 \leq j < n$. The function declaration is

vec lw(const double & c, const vec & p).

The vector p has dimension n and has positive elements that are less than 1. For $0 \leq j < n$, the probability that $X_j = 1$ is element j of p. The variable c is normally a small positive number used as in lw.pdf to remove very small probabilities from consideration in order to speed computation. If c is not positive, then the modified Lord-Wingersky algorithm used by lw.cpp reduces to the conventional algorithm. The probability mass function is provided by *lw*, a vector with $n + 1$ elements. For $0 \leq k \leq n$, element k of *lw* is the probability that $S = k$.

lwm.cpp

The function lwm.cpp finds the probability mass function of the sum S of n mutually independent random variables X_j, $0 \leq j <$ with integer values from 0 to $I_j - 1$ for an integer $I_j > 1$. The function declaration is

vec lwm(const double & c, const vector<vec> & p).

Here p has n members. For $0 \leq j < n$, member j of p is the vector *p[j]* with I_j nonnegative elements. The sum of these elements is 1, and element k, $0 \leq k < I_j$, of *p[j]* is the probability that $X_j = k$. The probability mass function is provided by *lwm*, a vector with $K = 1 + \sum_{j=1}^{n}(I_j - 1)$ elements. Element k of *lwm*, $0 \leq k < K$, is the probability that $S = k$. The variable c is normally a small positive number used as in lw.pdf to remove very small probabilities from consideration in order to speed computation. If c is not positive, then the modified algorithm used by lwm.cpp

reduces to the conventional generalization of the Lord-Wingersky algorithm to sums of independent multinomial variables.

Tools for Line Searches

The functions in this section facilitate line searches during function maximization. Throughout discussions in this section and in Functions related to the Newton-Raphson algorithm and Functions Related to Gradient Methods, the theoretical background and the definitions of η, γ_1, γ_2, and κ are found in convergence.pdf. For some positive integer p and nonempty open convex set O of p-dimensional vectors, a continuously differentiable real function *f.value* on O is to be maximized by an iterative algorithm with a starting value in O. It is assumed that, for some real a, the set A of members of O at which *f.value* is at least a is closed and bounded, and the sets A_0 of members of O at which *f.value* exceeds a is nonempty. The function *f.value* is assumed to be strictly pseudoconcave on A_0. The starting values for algorithms are assumed to be in A_0. The convention is adopted that *f.value* has value *NaN* at any p-dimensional vector not in O.

maxlinq2.cpp

The function *maxlinq2.cpp* provides a line search for maximization algorithms. Only function values and gradients are used when *order* is 1, but Hessian matrices are computed if *order* is greater than 1. The function declaration is

maxf2v maxlinq2(const int & order, const params & mparams, const vec & v,
const maxf2v & vary0, const function<f2v(const int &, const vec &)>f).

Here the definition of *maxf2v* is

struct maxf2v{vec locmax; double max; vec grad, mat hess;};,

vary0.locmax is the starting vector for the line search, *vary0.max* is the value of *f.value* at the starting vector, *maxlinq2.grad* is the gradient of *f.value* at *vary0.locmax*, *maxlinq2.hess*, if computed, is the gradient of *f.value* at *vary0.locmax*, and *maxlinq.locmax* is the approximate location of the maximum of *f.value* on the half-line that starts at *vary0.locmax* and has direction *v*, *maxlinq2.max* is the approximate maximum of *f.value* on the half-line, *maxlinq2.grad* is the gradient of *f.value* at *maxlinq.locmax*, and *maxlinq2.hess*, if computed, is the Hessian of *f.value* at *vary0.locmax*,

The definition of *params* is

struct params{bool print; int maxit; int maxits; double eta;
double gamma1; double gamma2; double kappa; double tol;}.

Here *mparams.print* is used for output of the iteration number and function value at the end of the iteration, *mparams.maxit* is the number of primary iterations, *mparams.maxits* is the maximum number of uses of maxquad.cpp permitted for each primary iteration, *mparams.eta* is η, *mparams.gamma1* is γ_1, *mparams.gamma2* is γ_2, and *mparams.kappa* is κ. Iterations cease if the function value changes less than *mparams.tol* after a primary iteration.

The definition of *f2v* is

struct f2v{double value; vec grad; vec hess};,

where *f.value* is the function value, *f.grad* is the gradient of *f.value*, and *f.hess* is the Hessian of *f.value*.

The functions maxf2vvar.cpp, maxquad.cpp, modit.cpp, and rebound.cpp are all used.

maxquad.cpp

The function maxquad.cpp approximates the maximum of *f.value* along a half-line by use of a quadratic two-point approximation. The function declaration is

double maxquad(const double & x0, const double & x1, const double & f0,
const double & f1, const double & g0, const double & stepmax).

Here *x0* and *x1* are the points used, *f0* is the function value at *x0*, *f1* is the function value at *x1*, *g0* is the derivative at *x0*, and *stepmax* is the maximum change from *x0* permitted in the estimated location *maxquad* of the function maximum.

modit.cpp

The function modit.cpp truncates an iteration to conform to limits on step size and bounds in the case of a real function of one variable with a unique critical point and a limit of $-\infty$ as the absolute value of the function argument approaches ∞. The function declaration is

double modit(const double & eta, const double & alpha0, const double & alpha1,
const double & stepmax, const bounds & b),

and the struct *bounds* is defined as

struct bounds {double lower; double upper;}.

Here *eta* corresponds to η, *alpha0* is the previous location, *alpha1* is the proposed new location, *stepmax* is the positive limit on step size, *b.lower* is the lower

bound, and *b.upper* is the upper bound. It is assumed that *alpha0* and *alpha1* are different. The function returns a value *modit* that is normally *alpha1*; however, if *alpha1* exceeds *alpha0*, then *modit* is truncated above so that it does not exceed the minimum of *alpha0+stepmax* and *alpha0+eta(b.upper-alpha0)*, while if *alpha1* is less than *alpha0*, then *modit* is truncated below so that it is at least the maximum of *alpha0-stepmax* and *alpha0+eta(b.lower-alpha0)*.

rebound.cpp

The function rebound.cpp updates the lower and upper bounds for maximization of a differentiable real function on the real line with a unique critical point and a limit of $-\infty$ as the absolute value of the function argument approaches ∞. The function declaration is

bounds rebound(const double & y, const double & der, const bounds & b).

The struct *bounds* is defined as in modit.cpp. Here *y* is the current location, *der* is the function derivative at *y*, *b.lower* is the current lower bound, and *b.upper* is the current upper bound. It is assumed that *der* is not 0. If *der* is positive, *modit.lower* is *y* and *modit.upper* is *b.upper*. If *der* is negative, *modit.upper* is *y* and *modit.lower* is *b.lower*.

Functions related to the Newton-Raphson algorithm

In this section, functions are discussed that are related to the Newton-Raphson algorithm. It should be noted that references to function values, gradients, and Hessian matrices do not address computational methods. In fact, the function values, gradients, and Hessian matrices employed may be approximations derived by numerical differentiation or large-sample approximations. In this section, *f.value* is assumed to be twice continuously differentiable.

maxf2vvar.cpp

The function maxf2vvar.cpp is used to combine information on a location and on a function's value, gradient, and Hessian matrix at the location. The function maxf2vvar.cpp has declaration

maxf2v maxf2vvar(const int & order, const vec & y, const f2v & fy);.

The structs *f2v* and *maxf2v* are defined as in maxlinq2.cpp. The returned value *maxf2vvar.locmax* is *y*, while *maxf2vvar.max* is *fy.value*, *maxf2var.grad* is *fy.grad*, and *maxf2var.hess* is *fy.hess* at *y*. If *order* is less than 1, only fy.value is considered If *order* is 1, *fy.value* and *fy.grad* are considered. If *order* exceeds 1, then *fy.value*, *fy.grad*, and *fy.hess* are used.

ngh.cpp

The function ngh.cpp finds gradients and Hessian matrices via numerical differentiation. The function declaration is

f2v ngh(const int & order, const double & delta, const vec & x,
const function <f2v(const int & , const vec &)>f).

If the variable *order* is 0, only the function value *f.value* at x at is reported. If *order* is 1, then *f.grad* at x is also reported by a numerical approximation with step size *delta*. If *order* is at least 2, then both *f.grad* and *f.hess* at x are reported by use of numerical approximations with step size *delta*. Values of *f.grad* are found by evaluation of *f.value* at vectors that differ from x in only one element. When an element of these vectors differ from those of x, the difference has absolute value *delta*. Values of *f.hess* are found by evaluation of *f.value* at vectors that differ from x in no more than two elements. When an element of these vectors differ from those of x, the difference has absolute value *delta*.

nrv.cpp

The function nrv.cpp applies a modified version of the Newton-Raphson algorithm to maximization of *f.value*. The function nrv.cpp has declaration

maxf2v nrv(const int & order, const params & mparams, const vec & start,
const function<f2v(const int &, vec &)> f).

The structs *f2v*, *maxf2v*, and *params* are defined as in maxlinq2.cpp. The starting vector *start* must be in O.

The function nrv.cpp uses maxf2vvar.cpp, maxlinq2.cpp, maxquad.cpp, modit.cpp, and rebound.cpp. The value of *order* should be at least 2.

nrvn.cpp

The function nrvn.cpp uses numerical differentiation for a modified version of the Newton-Raphson algorithm to maximization of *f.value*. The function nrvn.cpp has declaration

maxf2v nrvn(const int & order, const params & mparams, const vec & start,
const double & step, const function<f2v(const int &, vec &)> f).

The structs *f2v*, *maxf2v*, and *params* are defined as in maxlinq2.cpp. The starting vector *start* must be in O.

The function nrvn.cpp uses nrv.cpp, ngh.cpp, maxf2vvar.cpp, maxlinq2.cpp, maxquad.cpp, modit.cpp, and rebound.cpp. The value of *order* should be at least 2.

Functions Related to Gradient Methods

In this section, functions are considered based on gradient-based methods.

conjgrad.cpp

The function *conjgrad.cpp* implements a conjugate gradient algorithm for maximization of *f.value*. The function declaration is

maxf2v conjgrad(const int & order, const params & mparams,
const vec & start, const function<f2v(const int & , const vec &)> f).

The starting vector is *start*. The value of *order* must be at least 1. If *order* is at least 2, Hessian matrices are computed even though not used in the algorithm.

The function conjgrad.cpp uses maxf2vvar.cpp, maxlinq2.cpp, maxquad.cpp, modit.cpp, and rebound.cpp.

conjgradn.cpp

The function *conjgradn.cpp* uses numerical differentiation to implement a conjugate gradient algorithm for maximization of *f.value*. The function declaration is

maxf2v conjgradn(const int & order, const params & mparams,
const vec & start, const double & step, const function<f2v(const int & , const vec &)> f).

The starting vector is *start*. The value of *order* must be at least 1. If *order* is at least 2, Hessian matrices are computed even though not used in the algorithm. The positive step size for differentiation is *step*.

The function conjgradn.cpp uses conjgrad.cpp, ngh.cpp, maxf2vvar.cpp, maxlinq2.cpp, maxquad.cpp, modit.cpp, and rebound.cpp.

gradascent.cpp

The function gradascent.cpp uses a gradient-ascent algorithm for maximization of *f.value*. The function declaration for gradascent.cpp is

maxf2v gradascent(const order & , const params & mparams,
const vec & start, const function<f2v(const int & , const vec &)> f).

The functions maxf2vvar.cpp, maxlinq2.cpp, maxquad.cpp, modit.cpp, and rebound.cpp are used. Definitions are as in conjgrad.cpp.

gradascentn.cpp

The function gradascent.ncpp uses numerical differentiation to implement a gradient-ascent algorithm for maximization of *f.value*. The function declaration for gradascent.ncpp is

maxf2v gradascentn(const order & , const params & mparams,
const vec & start, const double & step, const function<f2v(const int & , const vec &)> f).

The functions gradascent.cpp, ngh.cpp, maxf2vvar.cpp, maxlinq2.cpp, maxquad.cpp, modit.cpp, and rebound.cpp are used. Definitions are as in conjgradn.cpp.

Log-likelihood Components

In this section, components of log-likelihood functions are provided. A component has the form $\ell_c(\boldsymbol{\beta}; \mathbf{Y}, A, F, q, r)$. Here the character A defines the type of model component involved, F is a distribution function with a positive and twice-continuously differential derivative F_1 such that $\log F_1$ has a negative second derivative. The integer $q > 0$ is the parameter dimension, and the integer $r > 0$ is the data dimension. The character A is in the set $\mathcal{A}$ with elements B(logit beta), C (cumulative), D (continuous), E (logit Dirichlet), G (graded), H (log gamma), L (multinomial logit), M (maximum of two independent Bernoulli variables), N (multivariate normal), P (log-mean Poisson case), R (rank logit), S (Bernoulli), and T (censored continuous). Distribution functions used in this section are in the set $\mathcal{F}$ with three members, G, the standard Gumbel distribution function with value $G(y) = \exp(-\exp(-y))$ for y real, Ψ, the standard logistic distribution function with value $\Psi(y) = 1/[1+\exp(-y)]$ for y real, and Φ, the standard normal distribution function with derivative $\Phi_1(y) = \exp(-y^2/2)/(2\pi)^{1/2}$ for real y. The value of F is only relevant in the cumulative, continuous, graded, Bernoulli, and censored continuous cases. The variables M, F, q, and r then define an open convex subset $O(A, F, q, r)$ of q-dimensional vectors and a set $\mathcal{Y}(A, F, q, r)$ of r-dimensional vectors. The vector $\boldsymbol{\beta}$ is in $O(A, F, q, r)$, and $\mathbf{Y}$ is in $\mathcal{Y}(A, F, q, r)$.

To treat both continuous and discrete log-likelihood components, the integral symbol $\int$ is used in the following sense. Consider a real function g on a nonempty finite-dimensional set C. If C is convex and has a nonempty interior and g is integrable, then $\int(g)$ denotes the integral of g over C. If C is finite or countably infinite and g is summable, then $\int(g)$ is the sum of $g(\mathbf{c})$ over $\mathbf{c}$ in C. More generally, let $\mathcal{D}$ be a finite or countably infinite collection of nonempty disjoint sets D that are either convex sets with nonempty interior or finite or countably-infinite sets. Let C be the union of the sets in $\mathcal{D}$, and let g_D denotes the restriction of g to D in $\mathcal{D}$. Let $\int(g_D)$ be defined for D in $\mathcal{D}$, and let the $\int(g_D)$, D in $\mathcal{D}$, be summable. Then $\int(g)$ is the sum of $\int(g_D)$ over

D in $\mathcal{D}$. Similar conventions apply if g is vector-valued or matrix-valued. The requirement is imposed here that, for $\boldsymbol{\beta}$ in $O(A, F, q, r)$, $\int(\exp(\ell_c(\boldsymbol{\beta}; \cdot, A, F, q, r))) = 1$. Here $\exp(\ell_c(\boldsymbol{\beta}; \cdot, A, F, q, r))$ is the function on $\mathcal{Y}(A, F, q, r)$ equal to $\exp(\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r))$ if $\mathbf{y}$ is in $\mathcal{Y}(A, F, q, r)$. The gradient function of $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ is $\nabla\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ and the corresponding Hessian matrix is $\nabla^2\ell_c(\cdot; \mathbf{y}, A, F, q, r)$.

For a positive integer n and an observation i, $0 \leq i < n$, positive integers q_i and r_i and character variables A_i in $\mathcal{A}$ and F_i in $\mathcal{F}$ are given. The component of the log likelihood for observation i involves the predicted random vector $\mathbf{Y}_i$ in $\mathcal{Y}(A_i, F_i, q_i, r_i)$, the q_i by p predicting matrix $\mathbf{X}_i$ in a nonempty set $\mathcal{X}_i$, the q_i-dimensional vector $\mathbf{o}_i$, and the positive real weight w_i. If $\boldsymbol{\tau}$ is in O, then let $\boldsymbol{\lambda}_i(\boldsymbol{\tau}) = \mathbf{o}_i + \mathbf{X}_i\boldsymbol{\tau}$ be in $O(A_i, F_i, q_i, r_i)$ for $0 \leq i < n$, and let the log-likelihood function under study have the form

$$\ell(\boldsymbol{\tau}) = \sum_{i=0}^{n-1} w_i \ell_c(\boldsymbol{\lambda}_i(\boldsymbol{\tau}); \mathbf{Y}_i, A_i, F_i, q_i, r_i). \tag{1}$$

It follows that the gradient of ℓ at $\boldsymbol{\tau}$ in O is

$$\nabla\ell(\boldsymbol{\tau}) = \sum_{i=0}^{n-1} w_i \mathbf{X}_i^T \nabla\ell_c(\boldsymbol{\lambda}_i(\boldsymbol{\tau}); \mathbf{Y}_i, A_i, F_i, q_i, r_i), \tag{2}$$

and the Hessian matrix of ℓ at $\boldsymbol{\tau}$ is

$$\nabla^2\ell(\boldsymbol{\tau}) = \sum_{i=0}^{n-1} w_i \mathbf{X}_i^T \nabla^2\ell_c(\boldsymbol{\lambda}_i(\boldsymbol{\tau}); \mathbf{Y}_i, A_i, F_i, q_i, r_i)\mathbf{X}_i. \tag{3}$$

The Hessian matrix $\nabla^2\ell(\boldsymbol{\tau})$ has the approximation

$$\tilde{\nabla}^2\ell(\boldsymbol{\tau}) = -\sum_{i=0}^{n-1} w_i \mathbf{X}_i^T \nabla\ell_c(\boldsymbol{\lambda}_i(\boldsymbol{\tau}); \mathbf{Y}_i, A_i, F_i, q_i, r_i)[\nabla\ell_i(\boldsymbol{\lambda}_i(\boldsymbol{\tau}); \mathbf{Y}_i, A_i, F_i, q_i)]^T \mathbf{X}_i \tag{4}$$

(Haberman, 2013; Louis, 1982) .

The functions $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ used are considered in this section. Some are examined in the literature on survival analysis (Cox, 1972; Kalbfleisch & Prentice, 2002), generalized linear models (McCullagh & Nelder, 1989), multivariate analysis (Anderson, 2003), and discrete choice (McFadden, 1973). It should be noted that names for models are somewhat variable in different references, especially for graded and cumulative cases. In addition, graded and cumulative cases are defined to be consistent with the Bernoulli cases. The following C++ functions are employed for common examples. The structs *f2v* are defined as in maxlinq2.cpp. If the argument *beta* is not in O_i, then all values returned equal *NaN*. It is assumed that the user of the function has verified that the input vector *y* is in $\mathcal{Y}_i$. In the cases under study in this section, unless otherwise stated, the components are strictly concave, so that ℓ is strictly concave whenever $\mathbf{X}_i$, $0 \leq i < n$, spans a space of dimension p. Conditions for a unique $\hat{\boldsymbol{\tau}}$ in O such that $\ell(\hat{\boldsymbol{\tau}})$ equals the supremum of ℓ over O are relatively

complex (Haberman, 1974, 1977, 1980). It is worth noting that in cases in which $\hat{\boldsymbol{\tau}}$ in O satisfies the conditions that $\nabla\ell(\hat{\boldsymbol{\tau}})$ is the p-dimensional vector $\mathbf{0}_p$ with all elements 0 and $\nabla^2\ell(\hat{\boldsymbol{\tau}})$ is negative definite, then O can be restricted to ensure that ℓ is strictly concave on O and $\hat{\boldsymbol{\tau}}$ is the only member of O such that $\ell(\hat{\boldsymbol{\tau}})$ equals the supremum of ℓ on O and, for $\boldsymbol{\tau}$ in O, $\nabla\ell(\boldsymbol{\tau})$ is only the vector with all elements 0 if $\boldsymbol{\beta}$ equals $\hat{\boldsymbol{\beta}}$. In all component functions, *order* is less than 1 if only the component value is computed, 1 if the component value and gradient are found, and greater than 1 if the component value, gradient, and Hessian are found. If *order* exceeds 2, the approximation of the Hessian by Equation 4 is employed. Repeated use is made of the struct *resp* with *vec* component *dresp* and *ivec* component *iresp*. In typical cases, *resp.dresp* or *resp.iresp* has no elements; however, exceptions do exist.

berresp.cpp

The function berresp.cpp is used to handle standard models for Bernoulli random variables. Here $q = r = 1$, A is S, $\mathcal{Y}(A, F, q, r)$ is the set of one-dimensional vectors $\mathbf{y}$ with $y(0)$ equal 0 or 1, and $O(A, F, q, r)$ is the set of all one-dimensional vectors, and F is in $\mathcal{F}$. For $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$ and $\boldsymbol{\beta}$ in $O(A, F, q, r)$,

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \begin{cases} \log(F(\beta(0)), & y(0) = 1, \\ \log(1 - F(\beta(0))), & y(0) = 0. \end{cases} \tag{5}$$

The function declaration is

f2v berresp(const int & order, const char & transform, const resp & y, const vec & beta).

If *transform* is G, then $F = G$, If *transform* is L, then $F = \Psi$. If *transform* is N, then $F = \Phi$. The function *berresp.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

The function berresp.cpp requires loglog.cpp, logit.cpp, and probit.cpp.

contresp.cpp

The function contresp.cpp computes the function value, gradient, and Hessian matrix associated with the distribution of a location and scale model for a continuous random vector. Here $r = 1$, $q = 2$, A is D, $\mathcal{Y}(A, F, q, r)$ is the set of all one-dimensional vectors, $O(A, F, q, r)$ is the set of all two-dimensional vectors $\boldsymbol{\beta}$ with element $\beta(1) > 0$, and F is in $\mathcal{F}$. For $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$ and $\boldsymbol{\beta}$ in $O(A, F, q, r)$,

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \log(\beta(1)) + \log(F_1(\beta(0) + \beta(1)y(0))). \tag{6}$$

These cases correspond to a model that a random variable Y has a distribution function $F(\beta(0) + \beta(1)y)$, where F is the distribution function of a random variable Z. Here $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ is concave, and the function is strictly concave if $y(0)$ is not 0.

For all cases, the function declaration is

f2v contresp(const int & order, const char & transform, const resp & y, const vec & beta).

The variable *transform* is defined as in berresp.cpp. The function *contresp.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.dresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. The function contresp.cpp requires gumbel.cpp, logistic.cpp, and normal.cpp.

cumresp.cpp

The function cumresp.cpp computes the function value, gradient, and Hessian matrix associated with a cumulative response transformation. Here $r = 1$, $q \geq 1$, A is C, F is in $\mathcal{F}$, $\mathcal{Y}(A, F, q, r)$ is the set of one-dimensional vectors $\mathbf{y}$ such that $y(0)$ is a nonnegative integer no greater than q, $O(A, F, q, r)$ is the set of all vectors of dimension q, and F is defined as in berresp.cpp. For $\boldsymbol{\beta}$ in $O(A, F, q, r)$ and $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$,

$$\ell_i(\boldsymbol{\beta}; \mathbf{y}) = \begin{cases} \log(1 - F(\beta(y(0))), & y(0) = 0, \\ \log(1 - F(\beta(y(0)))) + \sum_{i=0}^{y(0)-1} \log(F(\beta(i))), & 0 < y(0) < q, \\ \sum_{i=0}^{y(0)-1} \log(F(\beta(i))), & y(0) = q. \end{cases} \tag{7}$$

The function declaration is

f2v cumresp(const int & order, const char & transform, const resp & y, const vec & beta).

Here *transform* is defined as in berresp.cpp. The function *cumresp.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. The function cumresp.cpp requires berresp.cpp, loglog.cpp, logit.cpp, and probit.cpp. If $r = 1$, then use of cumresp.cpp is equivalent to use of berresp.cpp. In general, $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ is concave. Strict concavity holds if $q - y(0)$ does not exceed 1.

gradresp.cpp

The function gradresp.cpp computes the function value, gradient, and Hessian matrix associated with a graded response transformation. Then $r = 1$, $q \geq 1$, A is G, F is in $\mathcal{F}$, $O(A, F, q, r)$ is the set of all vectors of dimension q with strictly decreasing elements, $\mathcal{Y}(A, F, q, r)$ is the set of one-dimensional vectors $\mathbf{y}$ with $y(0)$ a nonnegative integer no greater than q, and, for $\boldsymbol{\beta}$ in $O(A, F, q, r)$ and $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$,

$$\ell_i(\boldsymbol{\beta}; \mathbf{y}) = \begin{cases} \log(1 - F(\beta(y(0))), & y(0) = 0, \\ \log(F(\beta(y(0) - 1)) - F(\beta(y(0)))), & 0 < y(0) < q, \\ \log(F(\beta(y(0) - 1))), & y(0) = q. \end{cases} \tag{8}$$

The function declaration is

f2v gradresp(const int & order, const char & transform, const resp & y, const vec & beta).

Here *transform* is defined as in berresp.cpp. The function *gradresp.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. If $q = 1$, then berresp.cpp, cumresp.cpp and gradresp.cpp yield the same result. The function $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ is concave. Strict concavity only holds if q is 1 or q is 2 and $y(0) = 1$.

gumbel.cpp

The function gumbel.cpp provides the computations required in contresp.cpp for $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ for the simple Gumbel case of $F = G$, A with value D, $q = 2$, $r = 1$, $\mathcal{Y}(A, F, q, r)$ the set of real numbers, and $O(A, F, q, r)$ the set of two-dimensional vectors $\boldsymbol{\beta}$ with $\beta(1) > 0$. The function declaration is

f2v gumbel(const int & order, const resp & y, const vec & beta).

The function *gumbel.value* is then $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.dresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

loggamma.cpp

The function loggamma.cpp provides the computations required for $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ for the log-gamma distribution with A with value H, $q = 2$, $r = 1$, $\mathcal{Y}(A, F, q, r)$ the set of real numbers, and $O(A, F, q, r)$ the set of two-dimensional vectors $\boldsymbol{\beta}$ with positive elements. The function declaration is

f2v loggamma(const int & order, const resp & y, const vec & beta).

The function *loggamma.value* is then $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.dresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

logistic.cpp

The function logistic.cpp provides the computations required in contresp.cpp for $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ for the logistic case $F = \Psi$ in contresp.cpp with A with value D, $q = 2$, $r = 1$, $\mathcal{Y}(A, F, q, r)$ the set of real numbers, and $O(A, F, q, r)$ the set of two-dimensional vectors $\boldsymbol{\beta}$ with $\beta(1) > 0$.. The function declaration is

f2v logistic(const int & order, const resp & y, const vec & beta).

The function *logistic.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.dresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

logit.cpp

The function logit.cpp computes the function value, gradient, and Hessian matrix associated with the logit case in berresp.cpp with A equal to S, $F = \Psi$, and $q = r = 1$. The function declaration is

f2v logit(const int & order, const resp & y, const vec & beta).

The function *logit.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

logitbeta.cpp

The function logitbeta.cpp computes the function value, gradient, and Hessian matrix associated with the logit of a beta distribution with a two-dimensional parameter vector $\boldsymbol{\beta}$ with positive elements. Here $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ has A with value H, $q = 2$, $r = 1$, $\mathcal{Y}(A, F, q, r)$ the set of real numbers, and $O(A, F, q, r)$ the set of two-dimensional vectors $\boldsymbol{\beta}$ with positive elements. The function declaration is

f2v logitbeta(const int & order, const resp & y, const vec & beta).

The function *logitbeta.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if $q = 2$, $r = 1$, *y.iresp* is $\mathbf{y}$, A is B, and *beta* is $\boldsymbol{\beta}$.

logitdirichlet.cpp

The function logitdirichlet.cpp computes the function value, gradient, and Hessian matrix associated with the logits of a Dirichlet distribution with a $q = r + 1$-dimensional parameter vector $\boldsymbol{\beta}$ with positive elements. Here $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ has A with value E, $\mathcal{Y}(A, F, q, r)$ the set of r-dimensional vectors, and $O(A, F, q, r)$ the set of q-dimensional vectors $\boldsymbol{\beta}$ with positive elements. The function declaration is

f2v logitdirichlet(const int & order, const resp & y, const vec & beta).

The function *logitdirichlet.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$, *beta* is $\boldsymbol{\beta}$, and A is E. The q-dimensional random variable $\mathbf{u}$ has a Dirichlet distribution with parameter vector $\boldsymbol{\beta}$ if 1 is the sum of the elements of $\mathbf{u}$ and $y_i = \log(u_i/u_q)$ for integers i from 1 to r. If $r = 1$, then the logit Dirichlet case reduces to the case of a logit beta.

loglog.cpp

The function loglog.cpp computes the function value, gradient, and Hessian matrix associated with the log-log case of berresp.cpp with A equal to S, $F = G$, and $q = r = 1$. The function declaration is

f2v loglog(const int & order, const resp & y, const vec & beta).

The function *loglog.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

logmean.cpp

The function logmean.cpp computes the function value, gradient, and Hessian matrix associated with a log-mean transformation for a Poisson random variable. In this case, $q = r = 1$, A is P, the value of F in $\mathcal{F}$ is irrelevant, $\mathcal{Y}(A, F, q, r)$ is the set of one-dimensional vectors $\mathbf{y}$ such that $y(0)$ is a nonnegative integer, and $O(A, F, Q, R)$ is the set of all one-dimensional vectors. For $\boldsymbol{\beta}$ in $O(A, F, q, r)$ and $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$,

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = y(0)\beta(0) - \exp(\beta(0)) - \log([y(0)]!). \tag{9}$$

The function declaration is

f2v logmean(const int & order, const resp & y, const vec & beta).

The function *logmean.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

maxberresp.cpp

The function maxberresp.cpp finds the log likelihood component, gradient, and Hessian matrix for the maximum of two unobserved Bernoulli random variables. Here $q = 2$, $r = 1$, A is M, F is in $\mathcal{F}$, $O(A, F, q, r)$ is the set of two-dimensional vectors, and $\mathcal{Y}(A, F, q, r)$ is the set of one-dimensional vectors $\mathbf{y}$ with $y(0)$ equal 0 or 1. For y in $\mathcal{Y}(A, F, q, r)$ and $\boldsymbol{\beta}$ in $O(A, F, q, r)$,

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \begin{cases} \log(F(\beta(0) + F(\beta(1) - F(\beta(0)F(\beta(1)), & y(0) = 1, \\ \log(1 - F(\beta(0))) + \log(1 - F(\beta(1))), & y(0) = 0. \end{cases} \tag{10}$$

It should be noted that

$$F(\beta(0) + F(\beta(1) - F(\beta(0)F(\beta(1)) = 1 - [1 - F(\beta(0))][1 - F(\beta(1)] \tag{11}$$

and

$$\log(1 - F(\beta(0)) + \log(1 - F(\beta(1)) = \log([1 - F(\beta(0)][1 - F(\beta(1)]). \tag{12}$$

The function $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ is not necessarily concave if $y(0) = 1$.

The function declaration is

f2v maxberresp(const int & order, const char & transform, const resp & y, const vec & beta).

The variable *transform* is defined as in berresp.cpp. The function *maxberesp.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.The functions berresp.cpp, logit.cpp, loglog.cpp, and probit.cpp are required.

multlogit.cpp

The function multlogit.cpp computes the function value, gradient, and Hessian matrix associated with a multinomial logit transformation. In this case, $r = 1$, $q \geq 1$, F is irrelevant, A is L, $\mathcal{Y}(A, F, q, r)$ is the set of one-dimensional vectors $\mathbf{y}$ such that $y(0)$ is a nonnegative integer no greater than q, and $O(A, F, q, r)$ is the set of all q-dimensional vectors. For $\boldsymbol{\beta}$ in $O(A, F, q, r)$ and $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$,

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \begin{cases} -\log\left(1 + \sum_{k=0}^{q-1} \exp(\beta(k)\right), & y(0) = 0, \\ \beta(y(0) - 1) + \ell_c(\boldsymbol{\beta}; \mathbf{0}_1, A, F, q, r), & y(0) > 0. \end{cases} \tag{13}$$

The function declaration is

f2v multlogit(const int & order, const resp & y, const vec & beta).

The function *multlogit.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. If $q = 1$, use of multlogit.cpp gives the same result as use of logit.cpp and as use of berresp.cpp, cumresp.cpp, or gradresp.cpp with *transform* equal L.

normal.cpp

The function normal.cpp computes the function value, gradient, and Hessian matrix associated with the normal case in contresp.cpp. Thus A is D, $F = \Phi$, $q = 2$, $r = 1$, $\mathcal{Y}(A, F, q, r)$ is the space of one-dimensional vectors, and $O(A, F, q, r)$ is the set of two-dimensional vectors $\boldsymbol{\beta}$ with $\beta(1) > 0$. The function declaration is

f2v normal(const int & order, const vec & y, const vec & beta).

The function *normal.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.dresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

normalv.cpp

The function normalv.cpp computes the function value, gradient, and Hessian matrix associated with the log-likelihood component associated with a multivariate normal model with r positive, $q = r(r + 3)/2$, A equal to N, F is irrelevant, $\mathcal{Y}(A, F, q, r)$ the set of all r-dimensional real vectors, and $O(A, F, q, r)$ the set of q-dimensional vectors $\boldsymbol{\beta}$ with elements β_h, $0 \leq h < q$ such that $\beta_h > 0$ if $h = r + j(j + 3)/2$ and $0 \leq j < r$. For such $\boldsymbol{\beta}$, let $\mathbf{a}(\boldsymbol{\beta})$ be the r-dimensional vector with elements $a_j(\boldsymbol{\beta}) = \beta_j$ for $0 \leq j < r$, and let $\mathbf{B}(\boldsymbol{\beta})$ be the symmetric positive-definite r by r matrx with row j and column k equal to β_h if $0 \leq k \leq j < r$ and $h = r + k + (j(j + 1)/2$. For an r-dimensional vector $\mathbf{z}$ with elements z_j, $0 \leq j < q$, let $\phi(\mathbf{z}; r)$ be the product of the $\Phi_1(z_j)$, $0 \leq j < r$.

For $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$,

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}) = \left[\sum_{j=0}^{r-1} \log(\beta(r + j(j+3)/2))\right] + \log(\phi(\mathbf{a}(\boldsymbol{\beta}) + \mathbf{B}(\boldsymbol{\beta})\mathbf{y}; r)). \tag{14}$$

This case corresponds to a model that a random vector has a distribution $\mathbf{a}(\boldsymbol{\beta}) + \mathbf{B}(\boldsymbol{\beta})\mathbf{Z}$, where $\mathbf{Z}$ is an r-dimensional multivariate normal random vector with zero mean and with covariance matrix equal to the identity matrix. The function $\ell_c(\cdot; \mathbf{y}, A, F, q, r)$ is always concave but is not strictly concave. The function declaration is

f2v normalv(const int & order, const resp & y, const vec & beta).

The function *normalv.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.dresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. If r is 1, then normalv.cpp reduces to normal.cpp. The function normalv.cpp requires pack.cpp and unpack.cpp.

pack.cpp

For the struct *vecmat* defined by

struct vecmax{vec v; mat m;};,

the function pack.cpp converts a d-dimensional vector *pack.v* and a d by d symmetric matrix *pack.m* to a vector with dimension $d(d+3)/2$ with d initial elements the vector *pack.v* and element $h = d + k + (j(j+1)/2$ equal to row j and column k of *pack.m* for nonnegative k no greater than $j < d$. The function declaration is

vec pack(const vecmat & u).

Diagonal elements of the matrix equal the corresponding elements of u, and off-diagonal elements are twice the corresponding elements of the vector.

probit.cpp

The function probit.cpp computes the function value, gradient, and Hessian matrix associated with a probit transformation in berresp.cpp with A equal to S, $F = \Psi$, and $q = r = 1$. The function declaration is

f2v probit(const int & order, const resp & y, const vec & beta).

The function *probit.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. $F = \Phi$. The function declaration is

f2v probit(const int & order, const resp & y, const vec & beta).

The function *probit.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$.

ranklogit.cpp

The function ranklogit.cpp computes the function value, gradient, and Hessian matrix associated with a model for discrete choice in which $q+1$ objects are ranked for some positive integer q and the r most-preferred objects are recorded for some positive integer $r \leq q$. Here A has value R, F is irrelevant, the set $\mathcal{Y}(A, F, q, r)$ consists of the vectors $\mathbf{y}$ of dimension r with distinct nonnegative integer elements that are no greater than q, and $O(A, F, q, r)$ is the set of all q-dimensional vectors. To describe the model, consider the standard Gumbel distribution function G. Consider $\boldsymbol{\beta}$ in $O(A, F, q, r)$. Let $U(j)$, $0 \leq j \leq q$, be independent random variables such that $U(0)$ and $U(j) - \beta(j)$, $1 \leq j \leq q$, have the common distribution function G. Let $\mathbf{Y}$ be a random vector with values in $\mathcal{Y}(A, F, q, r)$ such that $\mathbf{Y}$ is the member $\mathbf{y}$ of $\mathcal{Y}(A, F, q, r)$ with elements $y(j)$, $0 \leq j < r$, if $U(y(j))$ is nonincreasing in j and $U(y(j)) \geq U(k)$ if k is a nonnegative integer no greater than q that does not equal $y(h)$ for any nonnegative integer element $h < r$. For $\boldsymbol{\beta}$ in $O(A, F, q, r)$ and $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$, let $\boldsymbol{\alpha}(\boldsymbol{\beta})$ be the vector of dimension $q+1$ such that element j, $0 \leq j \leq q$, is $\alpha(j; \boldsymbol{\beta}) = 0$ if $j = 0$ and $\alpha(j; \boldsymbol{\beta}) = \beta(j-1)$ if $j > 0$. For $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$ and $0 \leq j < r$, let $K(j; \mathbf{y})$ be the set of nonnegative integers no greater than q not equal to $y(h)$ for any nonnegative integer $h < j$. Thus $K(0; \mathbf{y})$ is the set of nonnegative integers no greater than q. Then the log-likelihood component is

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \sum_{j=0}^{r-1} \left[\alpha(y_j; \boldsymbol{\beta}) - \log \left(\sum_{h \in K(j; \mathbf{y})} \exp(\alpha(h; \boldsymbol{\beta})) \right) \right]. \tag{15}$$

The function declaration is

f2v ranklogit(const int & order, const resp & y, const vec & beta).

The function *ranklogit.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$ if *y.iresp* is $\mathbf{y}$ and *beta* is $\boldsymbol{\beta}$. If $r = 1$, use of ranklogit.cpp gives the same result as use of multlogit.cpp.

truncresp.cpp

The function truncresp.cpp computes the function value, gradient, and Hessian matrix associated with a right-censored continuous random variable with the distribution of $\beta(0) + \beta(1)Z$ for some real $\beta(0)$ and positive real $\beta(1)$, where, as in contresp.cpp, Z has distribution function F in $\mathcal{F}$. In this case, $q = r = 2$, A is T. $\mathcal{Y}(A, F, q, r)$ consists of two-dimensional vectors $\mathbf{y}$ such that $y(0)$ is a real number and $y(1)$ is 0 or 1, and $O(A, F, q, r)$ is the set of all two-dimensional vectors $\boldsymbol{\beta}$ with

element $\beta(1) > 0$. For $\boldsymbol{\beta}$ in $O(A, F, q, r)$ and $\mathbf{y}$ in $\mathcal{Y}(A, F, q, r)$, if $y(1) = 0$, then the observation is not censored and the corresponding log-likelihood component is

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \log(\beta(1)) + \log(F_1(\beta(0) + \beta(1)y(0))), \tag{16}$$

while in the case of $y(1) = 1$, the the observation is censored at $y(0)$ and the log-likelihood component is

$$\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r) = \log(1 - F(\beta(0) + \beta(1)y(0))). \tag{17}$$

The function declaration is

f2v truncresp(const int & order, const char & transform, const resp & y, const vec & beta).

Here *y.iresp* has the single element $y(1)$, *y.dresp* has the single element $y(0)$, *beta* is $\boldsymbol{\beta}$, *transform* is defined as in berresp.cpp, and *truncresp.value* is $\ell_c(\boldsymbol{\beta}; \mathbf{y}, A, F, q, r)$. Functions required are berresp.cpp, contresp.cpp, and their respective required functions.

unpack.cpp

The function unpack.cpp converts a vector of dimension $d(d + 3)/2$ to the *vecmat* format described in pack.cpp. The function declaration is

vecmat unpack(const int & d, const vec & beta).

The vector *unpack.v* contains the first d elements of *beta* and row j and column k of *unpack.m* is element $d + k + j(j + 3)/2$ of *beta* for nonnegative integers $k \leq j < d$.

Computation of Log Likelihood Functions

genresp.cpp

The function genresp.cpp provides a general tool for computation of a a component of a log-likelihood function, its gradient, and its Hessian matrix. The function declaration is

f2v genresp(const int & order, const model & choice, const resp & y, const vec & beta).

Here *model* has the definition

struct model{char type; char transform}.

In *choice*, *choice.type* has value B for logit beta, C for a cumulative case, D for a continuous case, E for logit Dirichlet, G for a graded response, H for log gamma, L for the multinomial logit case, M for the maximum of two independent Bernoulli variables, N for the multivariate normal case, P for the log-mean Poisson case, R for the rank-logit case, S for the Bernouli case, and T for the censored continuous case. For discrete cases, *choice.transform* has possible values G for log-log cases, L for logit cases, and N for probit cases. For continuous cases, G is for the Gumbel distribution, L is for the logistic case, and N is for the normal case. For example, *choice.type* is C and *choice.transform* is G for the cumulative log-log case, while *choice.type* is S and *choice.type* is N in the probit case. The variable *choice.transform* is only relevant if *choice.type* is C, D, G, M, S, or T.

The function genresp.cpp uses berresp.cpp, contresp.cpp, cumresp.cpp, gradresp.cpp, loggamma.cpp, logitbeta.cpp, logitdirichlet.cpp, logmean.cpp, maxberresp.cpp, multlogit.cpp, ranklogit.cpp, and truncresp.cpp, together with the functions they in turn require.

genresplik.cpp

The function genresplik.cpp computes the log-likelihood function and its gradient and Hessian matrix. The function declaration is

f2v genresplik(const int & order, const vector<dat> & data, const xsel & obssel, const vec & beta).

The struct *dat* is defined by

struct dat{model choice; double weight; resp dep; vec offset; mat indep; xsel xselect;}.

Here *model* is defined as in genresp.cpp, *resp* is defined as in truncresp.cpp, and the struct *xsel* is defined by

struct xsel{bool all; uvec list}.

For $0 \leq i < n$, *data[i]* corresponds to observation i. Thus *data[i].choice* defines the model, *data[i].weight* is the observation weight w_i, *data[i].resp* defines the dependent vector $\mathbf{Y}_i$, *data[i].offset* is the offset vector $\mathbf{o}_i$, *data[i].indep* provides the matrix $\mathbf{X}_i$ of independent variables, and *data[i].xselect* is defined so that *x[i]* is $\mathbf{X}_i$ if *data[i].xselect[i].all* is *true*. Otherwise, two cases exist for $0 \leq j < p$. If *xselect[i].list* has K_i elements and j is *xselect[i].list(k)* for a nonnegative integer $k < K_i$, then column j of $\mathbf{X}_i$ is column k of *data[i].indep*. If j is not equal to any element of *x[i].list*, then column j of $\mathbf{X}_i$ is the q_i-dimensional vector with all elements 0. If *obssel.all* is

true, then all observations *data[i]* are used. Otherwise, *data[i]* is only used for i in *obssel.list*.

The function genresplik.cpp uses addsel.cpp, linsel.cpp, vecsel.cpp, genresp.cpp, and all C++ functions genresp.cpp requires.

genrespmle.cpp

The function genrespmle.cpp applies maximizes the log-likelihood function, gradient, and Hessian matrix of genresplik.cpp. The function declaration is

maxf2v genrespmle(const int & order, const params & mparams,
const char & algorithm, const vector<dat> & data, const xsel & obssel, const vec & start).

Here the structs *maxf2v* and *mparams* are defined as in maxlinq2.cpp and maxf2vvar.cpp. The vector *start* is the starting vector. The variable *algorithm* determines the algorithm, with N for Newton-Raphson, L for Newton-Raphson with the Hessian approximation of Equation 4, C for conjugate gradient, and G for gradient ascent. The functions nrv.cpp, conjgrad.cpp, gradascent.cpp, and genresplik.cpp are required, together with all C++ functions that these four functions need.

Tools for Computation of Log Likelihood Functions

addsel.cpp

The function addsel.cpp is used to add *f2v* structures. The function declaration is

void addsel(const int & order, const xsel & xselect,
const f2v & x, f2v & y, const double & a).

Here *order* and *xselect* are defined as in Log-likelihood Components. The struct y is modified by use of the struct x and the multiplier a. In all cases, *ax.value* is added to *y.value*. If *xselect.all* is *true*, then x and y have compatible dimensions, *ax.grad* is added to *y.grad* if *order* is at least 1, and *ax.hess* is added to *y.hess* if *order* is at least 2. If *xselect.all* is *false*, then *ax.grad* is added to *y.grad.elem(xselect.list)* if *order* is at least 1 and *ax.hess* is added to *y.hess.submat(xselect.list,xselect.list)* if *order* is at least 2.

ivecsel.cpp

The function ivecsel.cpp is employed to create a new integer vector from an old vector by extracting of elements of the old integer vector. The function declaration is

ivec ivecsel(const xsel & xselect, const ivec & y).

Here the struct *xsel* is defined as in genresplik.cpp. If *xselect.all* is *true*, then *ivecsel* is *y*. Otherwise, *ivecsel* is a vector with the number of elements in *xselect.list*, and element i of *ivecsel* is element *xselect.list(i)* of *y*.

linsel.cpp

The function linsel.cpp is used to apply a linear transformation to an *f2v* struct. The function declaration is

f2v linsel(const int & order, const f2v & x, const mat & a).

Here *order* is defined as in Log-likelihood Components. It is always the case that *linsel.value* is *x.value*. If *order* is positive, then *linsel.grad* is the product of the transpose of *a* and the gradient *x.grad*. If *order* exceeds 1, then *linsel.hess* is the product of the transpose of *a*, the Hessian *x.hess*, and the matrix *a*.

vecsel.cpp

The function vecsel.cpp is employed to create a new vector from an old vector by extracting of elements of the old vector. The function declaration is

vec vecsel(const xsel & xselect, const vec & y).

Here the struct *xsel* is defined as in genresplik.cpp. If *xselect.all* is *true*, then *vecsel* is *y*. Otherwise, *vecsel* is a vector with the number of elements in *xselect.list*, and element i of *vecsel* is element *xselect.list(i)* of *y*.

Latent Structures

In this section, functions useful for analysis of latent structures are considered. The log-likelihood function in this section is defined based on the definitions in Log-likelihood Components; however, use of latent variables is involved. In typical cases, data involve multiple responses for each individual observation. For a positive integer m, m observations are present. For observation h, $0 \leq h < m$, the observation has weight $w_{h*} > 0$, and n_h responses are observed. In addition, a latent vector appears in the model. Associated with the latent vector are positive integers q_* and r_*, A_* in $\mathcal{A}$, and F_* in $\mathcal{F}$. The latent vector $\boldsymbol{\theta}_h$ is in $\mathcal{Y}(A_*, F_*, q_*, r_*)$. The latent variable is predicted by the q_* by p predicting matrix $\mathbf{X}_{h*}$ in the nonempty set $\mathcal{X}_*$ and the fixed q_*-dimensional vector $\mathbf{o}_*$. It is assumed that $\boldsymbol{\lambda}_*(\boldsymbol{\tau}) = \mathbf{o}_* + \mathbf{X}_*\boldsymbol{\tau}$ is in $O(A_*, F_*, q_*, r_*)$ as long as $\mathbf{X}_*$ is in $\mathcal{X}_*$ and $\boldsymbol{\tau}$ is in O. For response i, $0 \leq i < n_h$, positive integers q_{hi} and r_{hi} are given. The variable A_{hi} is in $\mathcal{A}$ and F_{hi} is in $\mathcal{F}$. The component of the log likelihood for response i involves the predicted random vector

$\mathbf{Y}_{hi}$ in $\mathcal{Y}(A_{hi}, F_{hi}, q_{hi}, r_{hi})$, the latent vector $\boldsymbol{\theta}_h$, the q_{hi} by p predicting matrix $\mathbf{X}_{hi}$ in a nonempty set $\mathcal{X}_{hi}$, the q_{hi}-dimensional vector $\mathbf{o}_{hi}$, the q_{hi} by q_* matrix $\mathbf{D}_{hi}$, the positive real weight w_{hi}, the q_{hi} by p matrix $\mathbf{D}_{hik}$, $0 \leq k < q_*$, and the function $\ell_c(\cdot; \mathbf{y}, A_{hi}, F_{hi}, q_{hi}, r_{hi})$ on $O(A_{hi}, F_{hi}, q_{hi}, r_{hi})$ defined for $\mathbf{y}$ in $\mathcal{Y}(A_{hi}, F_{hi}, q_{hi}, r_{hi})$. For any $\boldsymbol{\tau}$ in O, $\mathbf{X}$ in $\mathcal{X}_{hi}$, and $\boldsymbol{\theta}$ in $\mathcal{Y}(A_*, F_*, q_*, r_*)$,

$$\boldsymbol{\lambda}_{hi}(\boldsymbol{\tau}|\boldsymbol{\theta}) = \mathbf{o}_{hi} + \mathbf{X}_{hi}\boldsymbol{\tau} + \mathbf{D}_{hi}\boldsymbol{\theta} + \sum_{k=0}^{p-1} \theta_k \mathbf{D}_{hik}\boldsymbol{\tau} \tag{18}$$

is in $O(A_{hi}, F_{hi}, q_{hi}, r_{hi})$.

For $\boldsymbol{\tau}$ in O, the log-likelihood has the form

$$\ell(\boldsymbol{\tau}) = \sum_{h=0}^{m-1} w_{h*} \ell_h(\boldsymbol{\tau}), \tag{19}$$

where $\ell_h(\boldsymbol{\tau})$ is the component of the log-likelihood for observation h. Thus the gradient function of ℓ at $\boldsymbol{\tau}$ satisfies

$$\nabla\ell(\boldsymbol{\tau}) = \sum_{h=0}^{m-1} w_{h*} \nabla\ell_h(\boldsymbol{\tau}), \tag{20}$$

where $\nabla\ell_h(\boldsymbol{\tau})$ is the gradient function of ℓ_h at $\boldsymbol{\tau}$. The Hessian function of ℓ at $\boldsymbol{\tau}$ satisfies

$$\nabla^2\ell(\boldsymbol{\tau}) = \sum_{h=0}^{m-1} w_{h*} \nabla^2\ell_h(\boldsymbol{\tau}), \tag{21}$$

where $\nabla^2\ell_h(\boldsymbol{\tau})$ is the Hessian function of ℓ_h at $\boldsymbol{\tau}$. The approximation

$$\tilde{\nabla}^2\ell(\boldsymbol{\tau}) = -\sum_{h=0}^{m-1} w_{h*} \nabla\ell_h(\boldsymbol{\tau})[\nabla\ell_h(\boldsymbol{\tau})]^T, \tag{22}$$

may also be considered.

In turn, $\ell_h(\boldsymbol{\tau})$ involves the product

$$\ell_h(\boldsymbol{\tau}|\boldsymbol{\theta}) = \ell_c(\boldsymbol{\lambda}_*(\boldsymbol{\tau}); \boldsymbol{\theta}, A_*, F_*, q_*, r_*) \sum_{i=0}^{n_h-1} w_{hi} \ell_c(\boldsymbol{\lambda}_{hi}(\boldsymbol{\tau}|\boldsymbol{\theta}); \mathbf{Y}_{hi}, A_{hi}, F_{hi}, q_{hi}, r_{hi}) \tag{23}$$

for $\boldsymbol{\theta}$ in $\mathcal{Y}(A_*, F_*, q_*, r_*)$. The component

$$\ell_h(\boldsymbol{\tau}) = \log \int (\exp(\ell_h(\boldsymbol{\tau}|\cdot))), \tag{24}$$

where $\exp(\ell_h(\boldsymbol{\tau}|\cdot))$ is the function with value $\exp(\ell_h(\boldsymbol{\tau}|\boldsymbol{\theta}))$ for $\boldsymbol{\theta}$ in $\mathcal{Y}(A_*, F_*, q_*, r_*)$. In practice, $\ell_h(\boldsymbol{\tau}))$ is evaluated by

$$\tilde{\ell}_h(\boldsymbol{\tau}) = \log \left[\sum_{k=1}^{Q} u_{hk} \exp(\ell_h(\boldsymbol{\tau}|\boldsymbol{\theta}_{hk})) \right], \tag{25}$$

for some positive weights u_{hk} and elements $\boldsymbol{\theta}_{hk}$ in $\mathcal{Y}(A_*, F_*, q_*, r_*)$.

irtc.cpp

The function irtc.cpp finds the conditional log likelihood component $\ell_h(\boldsymbol{\tau}|\boldsymbol{\theta})$ and associated gradient and Hessian matrix for a latent structure model. The function declaration is

f2v irtc (const int & order, const vector<dat> & data,
const vector<thetamap> & thetamaps, const resp & theta,
const xsel & datasel, const vec & beta).

In this declaration, *order* is less than 1 if only the function value is returned, at least 1 if the gradient is required, 2 if the Hessian is produced, and more than 2 if the approximate Hessian matrix is found. The use of *dat* is as in genresplik.cpp, the latent vector value is provided by *theta*, and *beta* is defined as usual as the parameter vector. Use of *datasel* is the same as use of *obssel* in genresplik.cpp.

The struct *thetamap* is defined by

struct thetamap{bool dep; xsel drespcols; xsel irespcols; mat offsets; cube indeps;}.

Here *dep* is true if the response is derived from *theta*, *drespcols* gives the elements of *theta.dresp* used in the response, and *irespcols* gives the elements of *theta.iresp* in the response. For item h and observation i, the matrix *thetamaps[i].offsets* contains each column k of $\mathbf{D}_{hi}$ such that k is specified in *thetamaps[i].drespcols*, and other columns of $\mathbf{D}_{hi}$ are zero vectors. In the cube *thetamaps[i].indeps*, each $\mathbf{D}_{hik}$ is given for k specified in *thetamaps[i].drespcols*, and column j of $\mathbf{D}_{hik}$ is provided if j is in the elements specified in *theta[i].drespcols* for item h and observation i. Other columns of $\mathbf{D}_{hik}$ are vectors of zeros. Element i of *data* corresponds to $\mathbf{Y}_{hi}$.

The functions ivecsel.cpp and genresplik.cpp, together with their associated functions, are required by irtc.cpp.

irtm.cpp

The function irtm.cpp finds the log likelihood component $\ell_h(\boldsymbol{\tau})$ and associated gradient and Hessian matrix for a latent structure model. The function uses numerical integration if $\mathcal{Y}(A_*, F_*, q_*, r_*)$ is not finite or countably infinite. The function declaration is

f2v irtm (const int & order, const vector<dat> & data,
const vector<thetamap> & thetamaps, const adq & scale, const params & mparamsn,
rescale & newscale,
const vector<pwr> & thetas, const xsel & datasel, const vec & beta).

In this declaration, only the function value is returned if *order* is 0, the function

value and gradient are returned if *order* is 1, and the function value, gradient, and Hessian matrix are returned if *order* is 2. The function value, gradient, and Louis approximation for the Hessian matrix are returned if *order* is 3. The use of *dat* is as in genresplik.cpp, and *beta* is defined as usual as the parameter vector. The struct *thetamap* is defined as in irtc.cpp. The definition of *datasel* is as in irtc.cpp.

The struct *pwr* is defined by

struct pwr{double weight; resp theta;},

the struct *adq* is

struct adq{bool adapt; xsel xselect; double step;},

and the struct *rescale*

struct rescale{double mult; vecmat tran},

The structs *scale* and *newscale* are important in adaptive quadrature (Naylor & Smith, 1982) and are used in irtmle.cpp to improve efficiency and accuracy of computations involving continuous latent vectors. Let *thetas* have Q elements *thetas[k]* such that *thetas[k].theta* corresponds to the q_*-dimensional vector $\boldsymbol{\theta}_k$. If *scale.adapt* is *false*, *scale.xselect.all* is *false* and *scale.xselect.list* has no elements, or *scale.xselect.all* is *true* and *thetas[0].theta.dresp* has no elements, then u_{hk} is *thetas[k].weight* and $\boldsymbol{\theta}_{hk}$ is $\boldsymbol{\theta}_k$. Consider the case of *scale.adapt* equal *true*, *thetas[0].theta.dresp* has a positive number of elements, and either *scale.xselect.all* is *true* or *scale.xselect.list* has a positive number of distinct nonnegative integer elements less than q_* such that each element of *scale.xselect.list* corresponds to an element of *thetas[0].theta.dresp*. Then $\boldsymbol{\theta}_{hk} = \mathbf{v}_h + \mathbf{M}_h\boldsymbol{\theta}_k$ for a q_*-dimensional vector $\mathbf{v}_h$ and a symmetric q_* by q_* matrix $\mathbf{M}_h$. An element of $\mathbf{v}_h$ is forced to be 0 if the corresponding element of $\boldsymbol{\theta}_0$ does not correspond to an element of *thetas[0].theta.dresp* and, in the case of *scale.xselect.all* equal *false*, does not correspond to an element of *thetas[0].theta.dresp* specified by *scale.xselect.list*. All members of a row of $\mathbf{M}_h$ are forced to be 0 if the corresponding element of $\mathbf{v}_h$ is forced to be 0. The elements of $\mathbf{v}_h$ not forced to be 0 are given in *rescale.tran.v*, and elements of $\mathbf{M}_h$ not forced to be 0 are given by *rescale.tran.m*. The matrix *rescale.tran.m* is positive-definite and *rescale.mult* is its determinant. If *rescale.tran.v* is used, then $\mathbf{v}_h$ is selected to maximize $\ell_h(\boldsymbol{\tau}|\boldsymbol{\theta})$ subject to the constraint that an element of $\boldsymbol{\theta}$ must equal the corresponding element of $\boldsymbol{\theta}_0$ when that element of $\mathbf{v}_h$ must be 0. Let this location of the maximum be $\hat{\boldsymbol{\theta}}_{h0}$. This maximization is normally accomplished with nrvn.cpp with step size *scale.mult*. The starting value is a previously specified *rescale.tran.v*. The matrix *rescale.tran.m* is normally the symmetric matrix square root of the inverse of the matrix formed from the rows and columns of the negative Hessian matrix of $\ell_h(\boldsymbol{\tau}|\cdot)$ at $\hat{\boldsymbol{\theta}}_{h0}$ that correspond to elements

of $\mathbf{v}_h$ not restricted to be 0, and *rescale.mult* is the corresponding determinant. If the desired symmetric square root does not exist, then *rescale* is unchanged from its previous value. The function irtm.cpp uses the functions irtc.cpp, adaptpwr.cpp, and nrvn.cpp, together with their required functions. The parameters in *mparamsn* are defined as in maxlinq2.cpp. They are used with nrvn.cpp to find *rescale.tran.v*.

irtmle.cpp

The function irtmle.cpp finds the maximum likelihood estimate for a latent structure model. The function uses numerical integration if $\mathcal{Y}(A_*, F_*, q_*, r_*)$ is not finite or countably infinite. The function declaration is

maxf2v irtmle (const int & order, const params & mparams,
const char & algorithm, const vec & obsweight,
const vector<vector<dat> > & obsdata,
const vector<vector<thetamap> > & obsthetamaps,
const vector<xsel> & datasel, const xsel & obssel,
const adq & scale, const params & mparamsn, vector <rescale> & obsscale,
const vector<pwr > & thetas, const vector<xsel> & betasel, const vec & start).

In this declaration, *maxf2v*, *mparams*, and *mparamsn* are defined as in maxlinq2.cpp and maxf2vvar.cpp, while *order* is defined as in irtc.cpp. The parameters in *mparamsn* are used when adaptive quadrature is specified, whereas *mparams* is used for the basic iterations used for determination of the maximum-likelihood estimate, The variable *algorithm* is defined as in genrespmle.cpp. The vector *obsweight* provides the weights w_{h*} for $0 \leq h \leq m-1$. The struct *dat* is defined as in genresplik.cpp, *obsdata[h]* is the data specification for observation h, the struct *thetamap* is defined as in irtc.cpp, *obsthetamap[h]* provides the mapping of the latent vector for observation h, the struct *pwr* is defined as in irtm.cpp, *thetas* is defined as in irtm.cpp, the struct *adq* is defined as in irtm.cpp, the scale adjustment for observation h is defined by *obsscale[h]*, the elements of *data[h]* to be used are defined as in *datasel[h]*, and the observations h used are defined by *obssel*. If *obssel.all* is *true*, then all observations are used. Otherwise, observation h only is used if in *obssel.list*. The vector *beta* is defined as usual as the parameter vector. The value of *betasel[h].all* is *true* if all elements of *beta* are used for the log likelihood for observation h, and the elements of *beta* corresponding to *betasel[h].list* are used for this log likelihood if *betasel[h].all* is *false*.

The functions irtms.cpp and its required functions and the functions that are prerequisites for genrespmle.cpp are required by irtmle.cpp.

irtms.cpp

The function irtms.cpp finds the log likelihood component $\ell(\boldsymbol{\tau})$ and associated gradient and Hessian matrix for a latent structure model. The function uses numerical integration if $\mathcal{Y}(A_*, F_*, q_*, r_*)$ is not finite or countably infinite. The function declaration is

f2v irtms (const int & order, const vec & obsweight,
const vector<vector<dat> > & obsdata,
const vector<vector<thetamap> > & obsthetamaps,
const vector<xsel> & datasel, const xsel & obssel, const adq & obsscale,
const params & mparamsn, vector<rescale>obsscales, const vector<pwr> & thetas,
const vector<xsel> & betasel, const vec & beta).

In this declaration, *order* is less than 1 if only the function value is returned, at least 1 if the gradient is required, 2 if the Hessian is produced, and more than 2 if the approximate Hessian matrix is found. The use of *obsweight*, *obsdata*, *obsthetamaps*, *datasel*, *obssel*, *scale*, *mparamsn*, *obsscales*, *thetas*, and *betasel* are as in irtmle.cpp, and *beta* is defined as usual as the parameter vector. The function irtm.cpp and its required functions are used by irtms.cpp.

posterior.cpp

For an observation, the function posterior.cpp finds the posterior distribution of the latent vector given the observed responses. The function declaration is

vecmat posterior (const vector<dat> & data,
const vector<thetamap> & thetamaps, const adq & scale, const rescale & newscale,
const vector<pwr> & thetas, const xsel & datasel, const vec & beta).

Definitions are consistent with those in irtm.cpp. The posterior is presented as q points, each of which has r elements. Output is based on *thetas*. The number q of points is the number of elements of *thetas*, and r is the number of elements of *thetas[0].theta.dresp*. The output consists of the vector *posterior.v* with q elements corresponding to the probabilities assigned to each point and the q by r matrix *posterior.m* with each row corresponding to a point of dimension r. All functions required by adaptpwr.cpp and irtc.cpp are required by posterior.cpp.

posteriors.cpp

For each observation used in irtms.cpp, the function posteriors.cpp finds the posterior distribution of the latent vector given the observed responses. The function declaration is

vector<vecmat> posteriors (const vector<vector<dat> > & obsdata,
const vector<vector<thetamap> >& obsthetamaps, const vector<xsel> & datasel,
const xsel obssel, const adq & scale, const vector<rescale> & obsscale,
const vector<pwr> & thetas, const vector<xsel> & betasel, const vec & beta).

Definitions are consistent with those in irtms.cpp. The posteriors for each observation are presented as in posterior.cpp. All functions required by posterior.cpp are required by posteriors.cpp.

Integration Tools

adaptpwr.cpp

The function adaptpwr.cpp provides a linear transformation of a quadrature point. The linear transformation has the form $L(\mathbf{x}) = \mathbf{a} + \mathbf{B}\mathbf{x}$ for the D-dimensional vector $\mathbf{x}$, where $\mathbf{a}$ is a D-dimensional vector and $\mathbf{B}$ is a D by D lower triangular matrix. The linear transformation is applied to each quadrature point and the weights are multiplied by the determinant of $\mathbf{B}$. The function declaration is

pwr adaptpwr(const pwr & oldtheta, const adq & scale, const rescale & newscale).

The structs *pwr*, *adq*, and *rescale* are defined as in irtm.cpp. If *scale.adapt* is *false*, *oldtheta.theta.dresp* has no elements, or *scale.xselect.all* is *false* and *scale.xselect.list* has no elements, then *adaptpwr* is *oldtheta*. In other cases, *adaptpwr.weight* is the product of *oldtheta.weight* and *newscale.mult*, *adaptpwr.theta.iresp* is *adaptpwr.theta.iresp*, elements of *adaptpwr.theta.dresp* that do not correspond to elements of *scale.tran.v* are the same as corresponding elements of *oldthat.theta.dresp*, and the remaining elements of *adaptpwr.theta.dresp* correspond to *rescale.tran.v* . plus the product of *rescale.tran.m* and elements of *oldtheta.theta.dresp* that correspond to elements of *rescale.tran.v*.

eaps.cpp

For each observation *i*, the function eaps.cpp generates a posterior weighted mean *eaps[i].v* and covariance matrix *eaps[i].m* that correspond to a discrete posterior distribution with points *posts[i].m.col(j)* with probabilities *posts[i].v(j)* for *j* from 1 to *posts[i].m.n_cols*. The function declaration is
vector<vecmat> eaps(const vector<vecmat> & posts).

The definition of *vecmat* is as in pack.cpp. The function wmc.cpp is used.

genfact.cpp

For a vector *sizes* of positive integers, the function genfact.cpp generates all vectors *i* of nonnegative integers with the same number of elements as *sizes* such

that each element of *i* is less than the corresponding element of *sizes*. The function declaration is

imat genfact(const ivec & sizes).

The columns of *genfact* are the possible vectors *i*. For example, if the elements of *sizes* are 2 and 3, then Column 0 of *genfact* has elements 0 and 0, and Column 1 has elements 1 and 0. In all, *sizes* has 6 columns, and Column 5 has elements 1 and 2.

genprods.cpp

The function genprods.cpp generates a collection of quadrature points and quadrature weights for a multivariate integral from quadrature weights and quadrature points for a univariate integral. The function declaration is

vecmat genprods(const imat & indices, const vector<pw> & pws).

The struct *vecmat* is defined as in pack.cpp, and the struct *pw* has *vec* elements *points* and *weights*. Consider the case of Q quadrature points for a multidimensional integral on the space of D-dimensional vectors, where Q and D are positive integers. Then *genprods.m* has Q columns and *genprods.v* has Q elements. The matrix *genprods.m* has D rows. The array *pws* has D members. For $0 \leq d < D$, *pws[d].points* and *pws[d].weights* have $m(d) > 1$ members, and the members of *pws[d].weights* are positive. The matrix *indices* specifies the quadrature vectors and quadrature weights to construct from *pws*. If *indices* has p columns, $0 \leq k < p$, and $0 \leq d < D$, then row d and column k of *indices* is nonnegative and less than $m(d)$ and the corresponding row and column of *genprods.m* is *pws[d].points(indices(d,k))*. Element k of *genprods.v* is the product of *pws[d].weights(indices(d,k))* for $0 \leq d < D$.

hermcoeff.cpp

The function hermcoeff.cpp finds the coefficients of a Hermite polynomial of a given degree. The function declaration is

vec hermcoeff(const int & n).

The integer variable n is the nonnegative order. The vector *hermcoeff* has n+1 elements. The polynomial is $H_n(x) = \sum_{i=0}^{n} \alpha_i x^{n-i}$ for real x, and element i of *hermcoeff* is α_i. For example, if n is 2, then the elements of *hermcoeff* are 1, 0, and -1.

hermpoly.cpp

The function hermpoly.cpp evaluates the Hermite polynomials up to a given degree at a specified real value. The function declaration is

vec hermpoly(const int &n, const double & x).

The degree is the nonnegative integer variable n, and the real value is x. The vector *hermpoly* has n+1 elements. For $0 \leq k \leq n$, element k of *hermpoly* is the value of H_k at x.

hermpw.cpp

The function hermpw.cpp uses the algorithm of Golub and Welsch (1969) to find the quadrature points and quadrature weights for Gauss-Hermite quadrature. The function declaration is

pw hermpw(const int & n).

The struct *hermpw* has vector elements *hermpw.points* and *hermpw.weights*. The number of quadrature points is n. The ordered quadrature points are in *hermpw.points*. The corresponding weights are in *hermpw.weights*. The weights are relative to the standard normal density.

normwt.cpp

The function normwt.cpp divides quadrature weights by the standard normal density to facilitate use with latent-structure models with latent variables that are normally distributed. The function declaration is

pw normwt(const pw & pwi).

The struct *normwt* has vector elements *normwt.points* and *normwt.weights*. The input *pwi* has elements *pwi.points* and *pwi.points*. The ordered quadrature points are in *normwt.points*. The corresponding weights are in *normwt.weights*. The result *normwt.points* is the same as *pwi.points*, but the weights *normwt.weights* are obtained from *pwi.weights* by dividing by the standard normal density at the corresponding points *pwi.points*.

qnormpw.cpp

The function *qnormpw.cpp* provides normal-scores quadrature of a given order. The function declaration is

pw qnormpw(const int & n).

The struct *qnormpw* has vector elements *qnormpw.points* and *qnormpw.weights*. The number of quadrature points is *n*. The ordered quadrature points are in *qnormpw.points*. The corresponding weights are in *qnormpw.weights*.

wmc.cpp

The function wmc.cpp computes a weighted mean vector and covariance matrix. The function declaration is

vecmat wmc(const vecmat & wx).

The elements of *wx.v* are probabilities corresponding to the rows of *wx.m*.

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