

<sup>+</sup>This function is part of [StataNow](#).

|                             |                          |                                      |                                |
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## Description

The `DerivDiscretePartial()` class computes discrete numerical partial derivatives at a list of points.

## Syntax

Syntax is presented under the following headings:

*Step 1: Initialization*

*Step 2: Definition of inputs and parameters*

*Step 3: Perform computation*

*Step 4: Display or obtain results*

*Definition of `q`*

*Functions defining inputs and parameters*

`q.setXValues()` and `q.getXValues()`

`q.setYValues()` and `q.getYValues()`

`q.setDirection()` and `q.getDirection()`

`q.setDegree()` and `q.getDegree()`

`q.setAccuracy()` and `q.getAccuracy()`

`q.setUseEqual()` and `q.getUseEqual()`

`q.setTol()` and `q.getTol()`

*Function for performing differentiation*

`q.solve()`

*Functions for obtaining results*

`q.getDerivatives()`

`q.converged()`

`q.errorcode()`, `q.errortext()`, and `q.returncode()`

### Step 1: Initialization

```
q = DerivDiscretePartial()
```

## Step 2: Definition of inputs and parameters

|                              |                                                               |
|------------------------------|---------------------------------------------------------------|
| <i>void</i>                  | <code>q.setXValues(pointer vector x)</code>                   |
| <i>void</i>                  | <code>q.setYValues(real vector data, real vector size)</code> |
| <i>void</i>                  | <code>q.setDirection(real vector direction)</code>            |
| <i>void</i>                  | <code>q.setDegree(real vector degree)</code>                  |
| <i>void</i>                  | <code>q.setAccuracy(real vector acc)</code>                   |
| <i>void</i>                  | <code>q.setUseEqual(real vector useeq)</code>                 |
| <i>void</i>                  | <code>q.setTol(real scalar tol)</code>                        |
| <i>pointer colvector</i>     | <code>q.getXValues()</code>                                   |
| <i>class NDMatrix scalar</i> | <code>q.getYValues()</code>                                   |
| <i>real colvector</i>        | <code>q.getDirection()</code>                                 |
| <i>real colvector</i>        | <code>q.getDegree()</code>                                    |
| <i>real colvector</i>        | <code>q.getAccuracy()</code>                                  |
| <i>real colvector</i>        | <code>q.getUseEqual()</code>                                  |
| <i>real scalar</i>           | <code>q.getTol()</code>                                       |

## Step 3: Perform computation

|             |                        |
|-------------|------------------------|
| <i>void</i> | <code>q.solve()</code> |
|-------------|------------------------|

## Step 4: Display or obtain results

|                              |                                 |
|------------------------------|---------------------------------|
| <i>class NDMatrix scalar</i> | <code>q.getDerivatives()</code> |
| <i>real scalar</i>           | <code>q.converged()</code>      |
| <i>real scalar</i>           | <code>q.errorcode()</code>      |
| <i>string scalar</i>         | <code>q.errortext()</code>      |
| <i>real scalar</i>           | <code>q.returncode()</code>     |

## Definition of q

A variable of type `DerivDiscretePartial()` is called an [instance](#) of the `DerivDiscretePartial()` class. `q` is an instance of `DerivDiscretePartial()`, a vector of instances, or a matrix of instances. If you are working interactively, you can create an instance of `DerivDiscretePartial()` by typing

```
q = DerivDiscretePartial()
```

For a row vector of  $n$  `DerivDiscretePartial()` instances, type

```
q = DerivDiscretePartial(n)
```

For an  $m \times n$  matrix of `DerivDiscretePartial()` instances, type

```
q = DerivDiscretePartial(m, n)
```

In a function, you would declare one instance of the `DerivDiscretePartial()` class  $q$  as a scalar.

```
void myfunc()
{
    class DerivDiscretePartial scalar    q
    q = DerivDiscretePartial()
    ...
}
```

You can instead declare  $q$  as a row vector of  $n$  instances by typing

```
void myfunc()
{
    class DerivDiscretePartial rowvector    q
    q = DerivDiscretePartial(n)
    ...
}
```

For an  $m \times n$  matrix of instances, type

```
void myfunc()
{
    class DerivDiscretePartial matrix    q
    q = DerivDiscretePartial(m, n)
    ...
}
```

## Functions defining inputs and parameters

At a minimum, you need to tell the `DerivDiscretePartial()` class about input vectors  $x$  and  $y$ .

Optionally, you may specify the directions, the degrees, and the accuracies of the derivatives. You may also specify whether the input vector  $x$  of points is equally spaced.

Each pair of functions includes a  $q.set*()$  function that specifies a setting and a  $q.get*()$  function that returns the current setting.

### **`q.setXValues()` and `q.getXValues()`**

$q.setXValues(x)$  sets the  $x$  values as a vector of pointers to real values. Each element in  $x$  points to a real vector that corresponds to a list of points (that is, points on the support of the function to be differentiated) sorted in ascending order with no duplicates.

$q.getXValues()$  returns the  $x$  values as a vector of pointers.

### **`q.setYValues()` and `q.getYValues()`**

$q.setYValues(data, size)$  sets the multidimensional  $y$  values in  $data$  as a vector and the lengths for each of the dimensions in  $size$  as a vector as well. Internally, these inputs will be stored as an instance of the `NDMatrix()` class.

If the actual dimension of the input matrix is  $n_1 \times n_2 \times \dots \times n_d$ , then the length of  $data$  is  $\prod_{i=1}^d n_i$ , and the value of  $size$  should be  $(n_1, n_2, \dots, n_d)$ .

`q.getYValues()` returns the instance of the `NDMatrix()` class that stores the multidimensional data and its size.

### **q.setDirection() and q.getDirection()**

`q.setDirection(direction)` sets the directions of differentiation as a vector indicating along which dimensions the derivative is to be taken. The default value for *direction* is 1.

`q.getDirection()` returns a vector of derivative directions.

### **q.setDegree() and q.getDegree()**

`q.setDegree(degree)` sets the degrees of differentiation as a vector indicating what the orders of the derivative are. The default value for *degree* is 1.

`q.getDegree()` returns a vector of the derivative degrees.

### **q.setAccuracy() and q.getAccuracy()**

`q.setAccuracy(acc)` sets the accuracies for differentiation as a vector. The default value for *acc* is 2.

`q.getAccuracy()` returns a vector of the derivative accuracies.

### **q.setUseEqual() and q.getUseEqual()**

`q.setUseEqual(useeq)` sets a vector of indicators for whether the input vectors of points are equally spaced. A value of 1 (the default) means yes and 0 means no.

Note that the class will rely on the values set in this function and will not check whether the input vectors are equally spaced.

`q.getUseEqual()` returns the setting for whether to assume an equally spaced vector of points.

### **q.setTol() and q.getTol()**

`q.setTol(tol)` specifies the computation tolerance. The default value of *tol* is 1e-8. See more details in [M-1] **Tolerance**.

`q.getTol()` returns the currently specified tolerance.

## **Function for performing differentiation**

### **q.solve()**

`q.solve()` computes the discrete numerical derivative based on the inputs.

## Functions for obtaining results

After performing the differentiation, the functions below return the derivatives, report convergence, and provide details on any errors.

### **q.getDerivatives()**

`q.getDerivatives()` returns the computed derivative as an instance of class `NDMatrix()`.

### **q.converged()**

`q.converged()` returns 1 if the computation converged and 0 if not.

### **q.errorcode(), q.errortext(), and q.returncode()**

`q.errorcode()` returns the error code generated during the computation; it returns 0 if no error is found.

`q.errortext()` returns an error message corresponding to the error code generated during the computation; it returns an empty string if no error is found.

`q.returncode()` returns the Stata return code corresponding to the error code generated during the computation.

The error codes and the corresponding Stata return codes are as follows:

| Error code | Return code | Error text                                       |
|------------|-------------|--------------------------------------------------|
| 1          | 430         | there are errors while computing weights         |
| 2          | 430         | dimensions do not conform                        |
| 3          | 111         | dimension of the input is 0                      |
| 4          | 430         | convergence is not achieved for some derivatives |

## Remarks and examples

Remarks are presented under the following headings:

[Introduction](#)  
[Examples](#)

### Introduction

The `DerivDiscretePartial()` class is a Mata class for computing discrete numerical derivatives.

Let  $x_i$  be a vector of length  $n_i$ , and let  $px_i$  be a pointer to the vector  $x_i$ , where  $i = 1, 2, \dots, d$ . Let  $y = f(x_1, x_2, \dots, x_d)$  be a multidimensional matrix representing some unknown function on  $(x_1, x_2, \dots, x_d)$ .

With this class, we can take any partial or mixed derivative of  $y$  with respect to some  $x_i$ 's.

At a minimum, you need to tell the `DerivDiscretePartial()` class about input vectors  $x$  and  $y$ . The argument for the method `q.setXValues()` should be a vector of pointers  $(px_1, px_2, \dots, px_d)$ , where  $px_i$  points to  $x_i$  for  $i = 1, 2, \dots, d$ . The arguments for the method `q.setYValues()` should be a vector *data* that represents  $y$  as a vector of length  $\prod_{i=1}^d n_d$  and a vector *size* that equals  $(n_1, n_2, \dots, n_d)$ .

For a vector of directions  $(j_1, j_2, \dots, j_k)$  and a vector of degrees  $(m_1, m_2, \dots, m_k)$ , the mixed partial derivative of  $y$  is

$$\frac{\partial^{\sum_{i=1}^k m_i} y}{\partial^{m_1} x_{j_1} \partial^{m_2} x_{j_2} \dots \partial^{m_k} x_{j_k}}$$

In this case, the argument for the method `q.setDirection()` should be a vector  $(j_1, j_2, \dots, j_k)$ , and the argument for the method `q.setDegree()` should be a vector  $(m_1, m_2, \dots, m_k)$ .

We may also set the accuracies  $(acc_1, acc_2, \dots, acc_k)$  for each part of the mixed partial derivatives using `q.setAccuracy()` and whether the input vectors are equally spaced (*useeq*<sub>1</sub>, *useeq*<sub>2</sub>, ..., *useeq*<sub>k</sub>) using `q.setUseEqual()`.

Note that if all the values of `q.setAccuracy()` are the same, then we can use a scalar instead of a vector as the argument. For example, if  $acc := acc_1 = acc_2 = \dots = acc_k$ , then we can use *acc* as the argument for `q.setAccuracy()`. This is also true for `q.setDirection()`, `q.setDegree()`, and `q.setUseEqual()`.

For an introduction to class programming in Mata, see [\[M-2\] class](#).

## Examples

To solve a discrete numerical derivative problem, you first use `DerivDiscretePartial()` to get an instance of the class. At a minimum, you must also use `q.setXValues()` and `q.setYValues()` to specify the input  $x$ 's and corresponding  $y$ 's.

### ► Example 1: Ordinary derivative

Consider the function  $f(x) = \sin(x)$ .

We first define an instance of the class:

```
: q = DerivDiscretePartial()
```

And then we set the input  $x$ 's and corresponding  $y$ 's. Remember that we also need to set the dimension of  $y$ .

```
: x = rangen(0, 1, 10)
: y = sin(x)
: q.setXValues(&x)
: q.setYValues(y, length(y))
```

Recall that the default direction is set to 1 and that the default degree of derivative is also set to 1. We assume the input  $x$  is equally spaced. We then set the computation accuracy to 5 and compute the derivative:

```
: q.setAccuracy(5)
: q.solve()
```

We check the result by typing

```
: q.getDerivatives().getData()
1
1      .999999336
2      .9938336501
3      .9754100103
4      .9449570257
5      .9028497782
6      .8496076999
7      .7858874242
8      .7124748124
9      .6302746686
10     .5403042494
```

Now consider an unequally spaced input  $x$ :

```
: x = rangen(0, 0.5, 5) \ rangen(0.51, 1, 10)
: y = sin(x)
: q.setXValues(&x)
: q.setYValues(y, length(y))
```

We need to indicate the input is not equally spaced and then compute the derivative by typing

```
: q.setUseEqual(0)
: q.solve()
: q.getDerivatives().getData()
1
1      .9999989701
2      .9921978752
3      .968912322
4      .9305077042
5      .877582566
6      .872744509
7      .8448859261
8      .814523552
9      .7817473667
10     .7466544989
11     .7093489463
12     .6699412626
13     .6285482313
14     .5852925003
15     .5403023675
```



## ► Example 2: Partial derivative

Consider the function  $f(x, y, z) = (x + y + z)^3 \exp(y^2) \log(z)$  and the problem of approximating  $(\partial y)/(\partial x)$  and  $(\partial^2 y)/(\partial y \partial z)$ .

We first define an instance of the class and set the input  $x$ 's and corresponding  $f$ 's. Remember that we also need to set the dimension of  $f$ .

```
: q = DerivDiscretePartial()
: x = rangen(-1, 2, 10)
: y = rangen(1, 2, 30)
: z = rangen(1, 1.5, 25) \ rangen(1.51, 2, 25)
: q.setXValues((&x, &y, &z))
: f = J(0, 1, 0)
: dfdx = J(0, 1, 0)
: d2fdydz = J(0, 1, 0)
: for (i = 1; i <= length(x); i++) {
>     for (j = 1; j <= length(y); j++) {
>         for (k = 1; k <= length(z); k++) {
>             f = f \ (x[i]+y[j]+z[k])^3 * exp(y[j]^2) * log(z[k])
>             dfdx = dfdx
>                 \ 3*(x[i]+y[j]+z[k])^2*exp(y[j]^2)*log(z[k])
>             d2fdydz = d2fdydz
>                 \ 6*(x[i]+y[j]+z[k])^2*y[j]*exp(y[j]^2)*log(z[k])
>                 + 6*(x[i]+y[j]+z[k])*exp(y[j]^2)*log(z[k])
>                 + 2*(x[i]+y[j]+z[k])^3*y[j]*exp(y[j]^2)/z[k]
>                 + 3*(x[i]+y[j]+z[k])^2*exp(y[j]^2)/z[k]
>         }
>     }
> }
: q.setYValues(f, (length(x), length(y), length(z)))
```

Here we also compute the exact solutions for  $(\partial y)/(\partial x)$  and  $(\partial^2 y)/(\partial y \partial z)$  so that we can compare them against our results.

To approximate  $(\partial y)/(\partial x)$ , recall that the default direction is set to 1 and that the default degree of differentiation is set to 1. We assume the input  $x$  is equally spaced. We then set the computation accuracy to 4 and compute the derivative:

```
: q.setAccuracy(4)
: q.solve()
```

We check the relative difference between our result and the exact solution:

```
: mrelldif(q.getDerivatives().getData(), dfdx)
1.58247e-14
```

Now consider the mixed derivative  $(\partial^2 y)/(\partial y \partial z)$ . We need to specify that one of the inputs is not equally spaced before we compute the derivative:

```
: q.setDirection((2, 3))
: q.setAccuracy(6)
: q.setUseEqual((1, 0))
: q.solve()
```

We again check the relative difference between our result and the exact solution:

```
: mrelldif(q.getDerivatives().getData(), d2fdydz)
7.94318e-06
```



## Conformability

`DerivDiscretePartial()`:

*input*:  
*output*:  
*result*:  $1 \times 1$   
*void*

`DerivDiscretePartial(n)`:

*input*:  
*n*:  $1 \times 1$   
*output*:  
*result*:  $1 \times n$

`DerivDiscretePartial(m, n)`:

*input*:  
*m*:  $1 \times 1$   
*n*:  $1 \times 1$   
*output*:  
*result*:  $m \times n$

`q.setXValues(x)`:

*input*:  
*x*:  $1 \times N$  or  $N \times 1$   
*output*:  
*result*: *void*

`q.getXValues()`:

*input*:  
*output*:  
*result*:  $N \times 1$   
*void*

`q.setYValues(data, size)`:

*input*:  
*data*:  $1 \times M$  or  $M \times 1$   
*size*:  $1 \times N$  or  $N \times 1$   
*output*:  
*result*: *void*

`q.getYValues()`:

*input*:  
*output*:  
*result*:  $1 \times 1$   
*void*

`q.setDirection(direction):`

*input:*

*direction:*  $1 \times N$  or  $N \times 1$  or  $1 \times 1$

*output:*

*result:* *void*

`q.getDirection():`

*input:*

*void*

*output:*

*result:*  $N \times 1$  or  $1 \times 1$

`q.setDegree(degree):`

*input:*

*degree:*  $1 \times N$  or  $N \times 1$  or  $1 \times 1$

*output:*

*result:* *void*

`q.getDegree():`

*input:*

*void*

*output:*

*result:*  $N \times 1$  or  $1 \times 1$

`q.setAccuracy(acc):`

*input:*

*acc:*  $1 \times N$  or  $N \times 1$  or  $1 \times 1$

*output:*

*result:* *void*

`q.getAccuracy():`

*input:*

*void*

*output:*

*result:*  $N \times 1$  or  $1 \times 1$

`q.setUseEqual(useeq):`

*input:*

*useeq:*  $1 \times N$  or  $N \times 1$  or  $1 \times 1$

*output:*

*result:* *void*

`q.getUseEqual():`

*input:*

*void*

*output:*

*result:*  $N \times 1$  or  $1 \times 1$

```

q.setTol(tol):
    input:
        tol:      1 × 1
    output:
        result:   void

q.getTol():
    input:
        result:   void
    output:
        result:   1 × 1

q.solve():
    input:
        result:   void
    output:
        result:   void

q.getDerivatives():
    input:
        result:   void
    output:
        result:   1 × 1

q.converged():
    input:
        result:   void
    output:
        result:   1 × 1

q.errorcode():
    input:
        result:   void
    output:
        result:   1 × 1

q.errortext():
    input:
        result:   void
    output:
        result:   1 × 1

q.returncode():
    input:
        result:   void
    output:
        result:   1 × 1

```

## Diagnostics

`DerivDiscretePartial()`, `q.set*()`, `q.get*()`, `q.solve()`, `q.converged()`, `q.errorcode()`, `q.errortext()`, and `q.returncode()` functions abort with an error message when used incorrectly.

## Also see

[M-5] **DerivDiscreteDiff()** — Finite difference coefficients for discrete numerical derivatives<sup>+</sup>

[M-4] **Mathematical** — Important mathematical functions

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