

**cvpermute()** — Obtain all permutations

[Description](#)    [Syntax](#)    [Remarks and examples](#)    [Conformability](#)  
[Diagnostics](#)    [Also see](#)

## Description

`cvpermute()` returns all permutations of the values of column vector  $V$ , one at a time. If  $V = (1\ 2\ 3)$ , there are six permutations and they are  $(1\ 2\ 3)$ ,  $(1\ 3\ 2)$ ,  $(2\ 1\ 3)$ ,  $(2\ 3\ 1)$ ,  $(3\ 1\ 2)$ , and  $(3\ 2\ 1)$ . If  $V = (1\ 2\ 1)$ , there are three permutations and they are  $(1\ 1\ 2)$ ,  $(1\ 2\ 1)$ , and  $(2\ 1\ 1)$ .

Vector  $V$  is specified by calling `cvpermutesetup()`,

```
info = cvpermutesetup(V)
```

`info` holds information that is needed by `cvpermute()` and it is `info`, not  $V$ , that is passed to `cvpermute()`. To obtain the permutations, repeated calls are made to `cvpermute()` until it returns `J(0,1,.)`:

```
info = cvpermutesetup(V)
while ((p=cvpermute(info)) != J(0,1,.)) {
    ... p ...
}
```

Column vector  $p$  will contain a permutation of  $V$ .

`cvpermutesetup()` may be specified with one or two arguments:

```
info = cvpermutesetup(V)
info = cvpermutesetup(V, unique)
```

`unique` is usually not specified. If `unique` is specified, it should be 0 or 1. Not specifying `unique` is equivalent to specifying `unique = 0`. Specifying `unique = 1` states that the elements of  $V$  are unique or, at least, are to be treated that way.

When the arguments of  $V$  are unique—for instance,  $V = (1\ 2\ 3)$ —specifying `unique = 1` will make `cvpermute()` run faster. The same permutations will be returned, although usually in a different order.

When the arguments of  $V$  are not unique—for instance,  $V = (1\ 2\ 1)$ —specifying `unique = 1` will make `cvpermute()` treat them as if they were unique. With `unique = 0`, there are three permutations of  $(1\ 2\ 1)$ . With `unique = 1`, there are six permutations, just as there are with  $(1\ 2\ 3)$ .

## Syntax

```
info = cvpermutesetup(real colvector V [ , real scalar unique ] )
real colvector cvpermute(info)
```

where *info* should be declared *transmorphic*.

## Remarks and examples

[stata.com](http://www.stata.com)

### ► Example 1

You have the following data:

| v1 | v2 |
|----|----|
| 22 | 29 |
| 17 | 33 |
| 21 | 26 |
| 20 | 32 |
| 16 | 35 |

You wish to do an exact permutation test for the correlation between *v1* and *v2*.

That is, you wish to (1) calculate the correlation between *v1* and *v2*—call that value *r*—and then (2) calculate the correlation between *v1* and *v2* for all permutations of *v1*, and count how many times the result is more extreme than *r*.

For the first step,

```
: X = (22, 29 \
>      17, 33 \
>      21, 26 \
>      20, 32 \
>      16, 35)
:
: correlation(X)
[symmetric]
           1           2
1  1
2  -.8468554653      1
```

The correlation is  $-.846855$ . For the second step,

```
: V1 = X[,1]
: V2 = X[,2]
: num = den = 0
: info = cvpermutesetup(V1)
: while ((V1=cvpermute(info)) != J(0,1,..)) {
>   rho = correlation((V1,V2))[2,1]
>   if (rho<=-.846 | rho>=.846) num++
>   den++
> }
```

```

: (num, den, num/den)
      1          2          3
1  ┌──────────┬──────────┬──────────┐
   │          13          120  .1083333333 │
   └──────────┴──────────┴──────────┘

```

Of the 120 permutations, 13 (10.8%) were outside .846855 or  $-.846855$ .



## ▶ Example 2

You now wish to do the same thing but using the Spearman rank-correlation coefficient. Mata has no function that will calculate that, but Stata has a command that does—see [R] [spearman](#)—so we will use the Stata command as our subroutine.

This time, we will assume that the data have been loaded into a Stata dataset:

```

. list

```

|    | var1 | var2 |
|----|------|------|
| 1. | 22   | 29   |
| 2. | 17   | 33   |
| 3. | 21   | 26   |
| 4. | 20   | 32   |
| 5. | 16   | 35   |

For the first step,

```

. spearman var1 var2
Number of obs =      5
Spearman's rho =  -0.9000
Test of H0: var1 and var2 are independent
Prob > |t| =      0.0374

```

For the second step,

```

. mata
----- mata (type end to exit) -----
: V1 = st_data(., "var1")
: info = cvpermutesetup(V1)
: num = den = 0
: while ((V1=cvpermute(info)) != J(0,1,.)) {
>   st_store(., "var1", V1)
>   stata("quietly spearman var1 var2")
>   rho = st_numscalar("r(rho)")
>   if (rho<=-.9 | rho>=.9) num++
>   den++
> }
: (num, den, num/den)
      1          2          3
1  ┌──────────┬──────────┬──────────┐
   │          2          120  .0166666667 │
   └──────────┴──────────┴──────────┘

```

Only two of the permutations resulted in a rank correlation of at least .9 in magnitude.

In the code above, we obtained the rank correlation from `r(rho)` which, we learned from [R] `spearman`, is where `spearman` stores it.

Also note how we replaced the contents of `var1` by using `st_store()`. Our code leaves the dataset changed and so could be improved.

◀

## Conformability

`cvpermutesetup(V, unique)`:

*V*:  $n \times 1$   
*unique*:  $1 \times 1$  (optional)  
*result*:  $1 \times L$

`cvpermute(info)`:

*info*:  $1 \times L$   
*result*:  $n \times 1$  or  $0 \times 1$

where

$$L = \begin{cases} 3 & \text{if } n = 0 \\ 4 & \text{if } n = 1 \\ (n + 3)(n + 2)/2 - 6 & \text{otherwise} \end{cases}$$

The value of  $L$  is not important except that the *info* vector returned by `cvpermutesetup()` and then passed to `cvpermute()` consumes memory. For instance,

| <i>n</i> | <i>L</i> | Total memory ( $8 \times L$ ) |
|----------|----------|-------------------------------|
| 5        | 22       | 176 bytes                     |
| 10       | 72       | 576                           |
| 50       | 1,372    | 10,560                        |
| 100      | 5,247    | 41,976                        |
| 1,000    | 502,497  | 4,019,976                     |

## Diagnostics

`cvpermute()` returns `J(0, 1, .)` when there are no more permutations.

## Also see

[M-4] **Statistical** — Statistical functions