

mi import flong — Import flong-like data into mi

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Description

`mi import flong` imports flong-like data, that is, data in which $m = 0, m = 1, \dots, m = M$ are all recorded in one `.dta` dataset.

`mi import flong` converts the data to mi flong style. The data are mi `set`.

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Syntax

```
mi import flong, required_options [true_options]
```

| <i>required_options</i> | Description |
|---------------------------------|---------------------------------|
| <code>m(<i>varname</i>)</code> | name of variable containing m |
| <code>id(<i>varlist</i>)</code> | identifying variable(s) |

| <i>true_options</i> | Description |
|--------------------------------------|--|
| <code>imputed(<i>varlist</i>)</code> | imputed variables to be registered |
| <code>passive(<i>varlist</i>)</code> | passive variables to be registered |
| <code>clear</code> | okay to replace unsaved data in memory |

Options

`m(varname)` and `id(varlist)` are required. `m(varname)` specifies the variable that takes on values $0, 1, \dots, M$, the variable that identifies observations corresponding to $m = 0, m = 1, \dots, m = M$. $varname = 0$ identifies the original data, $varname = 1$ identifies $m = 1$, and so on.

`id(varlist)` specifies the variable or variables that uniquely identify observations within `m()`.

`imputed(varlist)` and `passive(varlist)` are truly optional options, although it would be unusual if `imputed()` were not specified.

`imputed(varlist)` specifies the names of the imputed variables.

`passive(varlist)` specifies the names of the passive variables, if any.

`clear` specifies that it is okay to replace the data in memory even if they have changed since they were saved to disk. Remember, `mi import flong` starts with flong-like data in memory and ends with mi flong data in memory.

Remarks and examples

The procedure to convert flong-like data to mi flong is this:

1. use the unset data.
2. Issue the `mi import flong` command.
3. Perform the checks outlined in *Using mi import nhanes1, ice, flong, and flongsep* of **[MI] mi import**.
4. Use `mi convert` (see **[MI] mi convert**) to convert the data to a more convenient style, such as wide or mlong.

For instance, you have the following unset data:

```
. use http://www.stata-press.com/data/r15/ourunsetdata
(mi prototype)
. list, separator(2)
```

| | m | subject | a | b | c |
|----|---|---------|---|-----|-----|
| 1. | 0 | 101 | 1 | 2 | 3 |
| 2. | 0 | 102 | 4 | . | . |
| 3. | 1 | 101 | 1 | 2 | 3 |
| 4. | 1 | 102 | 4 | 4.5 | 8.5 |
| 5. | 2 | 101 | 1 | 2 | 3 |
| 6. | 2 | 102 | 4 | 5.5 | 9.5 |

You are told that these data contain the original data ($m = 0$) and two imputations ($m = 1$ and $m = 2$), that variable `b` is imputed, and that variable `c` is passive and in fact equal to $a + b$. These are the same data discussed in **[MI] styles** but in unset form.

The fact that these data are nicely sorted is irrelevant. To import these data, type

```
. mi import flong, m(m) id(subject) imputed(b) passive(c)
```

These data are short enough that we can list the result:

```
. list, separator(2)
```

| | m | subject | a | b | c | _mi_m | _mi_id | _mi_miss |
|----|---|---------|---|-----|-----|-------|--------|----------|
| 1. | 0 | 101 | 1 | 2 | 3 | 0 | 1 | 0 |
| 2. | 0 | 102 | 4 | . | . | 0 | 2 | 1 |
| 3. | 1 | 101 | 1 | 2 | 3 | 1 | 1 | . |
| 4. | 1 | 102 | 4 | 4.5 | 8.5 | 1 | 2 | . |
| 5. | 2 | 101 | 1 | 2 | 3 | 2 | 1 | . |
| 6. | 2 | 102 | 4 | 5.5 | 9.5 | 2 | 2 | . |

We will now perform the checks outlined in *Using mi import nhanes1, ice, flong, and flongsep* of **[MI] mi import**, which are to run `mi describe` and `mi varying` to verify that variables are registered correctly:

```
. mi describe
Style:  flong
      last mi update 21jan2017 12:52:18, 1 second ago
Obs.:  complete      1
      incomplete     1  (M = 2 imputations)
      -----
      total          2
Vars.:  imputed:    1; b(1)
      passive:     1; c(1)
      regular:     0
      system:      3; _mi_m _mi_id _mi_miss
      (there are 3 unregistered variables; m subject a)

. mi varying
```

| Possible problem | variable names |
|------------------------------|----------------|
| imputed nonvarying: | (none) |
| passive nonvarying: | (none) |
| unregistered varying: | (none) |
| *unregistered super/varying: | (none) |
| unregistered super varying: | m |

* super/varying means super varying but would be varying if registered as imputed; variables vary only where equal to soft missing in $m=0$.

We discover that unregistered variable `m` is [super varying](#) (see [\[MI\] Glossary](#)). Here we no longer need `m`, so we will drop the variable and rerun `mi varying`. We will find that there are no remaining problems, so we will convert our data to our preferred wide style:

```
. drop m
. mi varying
```

| Possible problem | variable names |
|------------------------------|----------------|
| imputed nonvarying: | (none) |
| passive nonvarying: | (none) |
| unregistered varying: | (none) |
| *unregistered super/varying: | (none) |
| unregistered super varying: | (none) |

* super/varying means super varying but would be varying if registered as imputed; variables vary only where equal to soft missing in $m=0$.

```
. mi convert wide, clear
. list
```

| | subject | a | b | c | _mi_miss | _1_b | _1_c | _2_b | _2_c |
|----|---------|---|---|---|----------|------|------|------|------|
| 1. | 101 | 1 | 2 | 3 | 0 | 2 | 3 | 2 | 3 |
| 2. | 102 | 4 | . | . | 1 | 4.5 | 8.5 | 5.5 | 9.5 |

Also see

[\[MI\] intro](#) — Introduction to mi

[\[MI\] mi import](#) — Import data into mi