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A common problem in data management is combining two or more variables with missing values to get a single variable with as many nonmissing values as possible. Typically, this problem arises when what should be the same variable has been named differently in different datasets. Stata’s treatment of missing values means that the combination needs a little care, although there are several quite easy solutions.

To make this situation concrete, consider a recent experience of mine. Two datasets were created. A variable `resp_ill` was created in one dataset, and the corresponding variable `moxResp` was created in the other. After appending the data with the `append` command, there were missing values on `resp_ill` if the data came from the second dataset, and there were missing values on `moxResp` if the data came from the first dataset. The total number of observations is 782. Here is a short listing of some of the observations:

```stata
. list if inrange(_n, 1, 5) | inrange(_n, 340, 344), separator(0)

<table>
<thead>
<tr>
<th>resp_ill</th>
<th>moxResp</th>
</tr>
</thead>
<tbody>
<tr>
<td>no</td>
<td>.</td>
</tr>
<tr>
<td>no</td>
<td>.</td>
</tr>
<tr>
<td>no</td>
<td>.</td>
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<td>no</td>
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<td>.</td>
<td>no</td>
</tr>
<tr>
<td>.</td>
<td>no</td>
</tr>
<tr>
<td>.</td>
<td>yes</td>
</tr>
</tbody>
</table>
```

The variables are indicators of whether the patient had a respiratory illness. They have been labeled as “no” for 0 and “yes” for 1.

It is easy to see or to say that the variables should be renamed (with the `rename` command) for consistency. I could have done that in one of the original datasets and then repeated the `append` command. In practice, it is often quicker to fix the problem in the current dataset. Sometimes, maintaining dataset integrity is sufficiently important to rule out the `rename` solution altogether.

Experimentation will confirm that adding the two variables is not the answer, because the missing values are propagated: for Stata, nonmissing plus missing is still missing. Doing that with a `replace` of one variable rather than a `generate` would be
an even worse idea because any original nonmissing values would be lost in the variable that was replaced.

What we want is to ignore the missings, and there are several ways to do that. One explicit solution is

```
. generate newind = min(resp_ill, moxResp)
```

The pairwise minimum will be whichever nonmissing value is present. What is less intuitive is that the function `max()` will work in the same way because it follows the same principle of ignoring missings to the extent possible, just as, say, `summarize` returns the maximum nonmissing value and not missings. So even though Stata has a general rule that missings count higher than nonmissings, in the case of the `max()` function, that rule is trumped by the principle that missings are ignored as far as possible.

Another solution is

```
. generate newind = cond(missing(resp_ill), moxResp, resp_ill)
```

In this particular case, it was known that the missings in one variable corresponded to the nonmissings in the other. In other circumstances, it would be prudent to check whether there were observations with nonmissing values on both variables, say, by typing

```
. list if var1 < . & var2 < .
```

Another solution that you may know is to use `egen`'s `rowtotal()` function, which by default ignores missings in forming a row total and so gives a single nonmissing result for one missing and one nonmissing argument. Using `egen` can be convenient, especially if you have more than two variables to combine, but it does impose an extra overhead because there is more code for Stata to interpret. For one-off tasks and small- or moderate-size datasets, you would have to strain to detect that overhead, but for repeated data management and large datasets, using `min()` or `max()` is preferable.