

## Resultssets, resultsspreadsheets and resultsplots in Stata

Roger Newson

Imperial College London, London, UK

[r.newson@imperial.ac.uk](mailto:r.newson@imperial.ac.uk)

<http://www.imperial.ac.uk/nhli/r.newson/>

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This presentation, and the do-files containing the examples, can be downloaded  
from the conference website at

<http://ideas.repec.org/s/boc/dsug06.html>

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- The tables may be in .pdf, .html, word processor, spreadsheet or paper documents.
- The plots may be produced using Stata or non–Stata software.
- This presentation will introduce ways of producing these plots and tables in Stata.

## **Results (1): Countries of origin of cars in the auto data**

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```
. tab country, gene(c_)
```

Country of origin of firm	Freq.	Percent	Cum.
-----+-----			
US	52	70.27	70.27
Japan	11	14.86	85.14
Germany	7	9.46	94.59
France	2	2.70	97.30
Italy	1	1.35	98.65
Sweden	1	1.35	100.00
-----+-----			
Total	74	100.00	

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weight	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
c_1	3317.115	88.26823	37.58	0.000	3140.979 3493.252
c_2	2248.182	191.9154	11.71	0.000	1865.221 2631.143
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Very few end users will understand this output!

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This resultsset is a Stata dataset created using `xcontract`, `parmest`, `descsave` and `factext` (downloadable from SSC). It has one observation per country, and data on frequencies, and on estimates and 95% confidence limits for mean weights (in US pounds), for cars made by firms based in that country.

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```
. list country _freq estimate min95 max95, clean noobs
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country	_freq	estimate	min95	max95
US	52	3317.12	3140.98	3493.25
Japan	11	2248.18	1865.22	2631.14
Germany	7	2238.57	1758.50	2718.64
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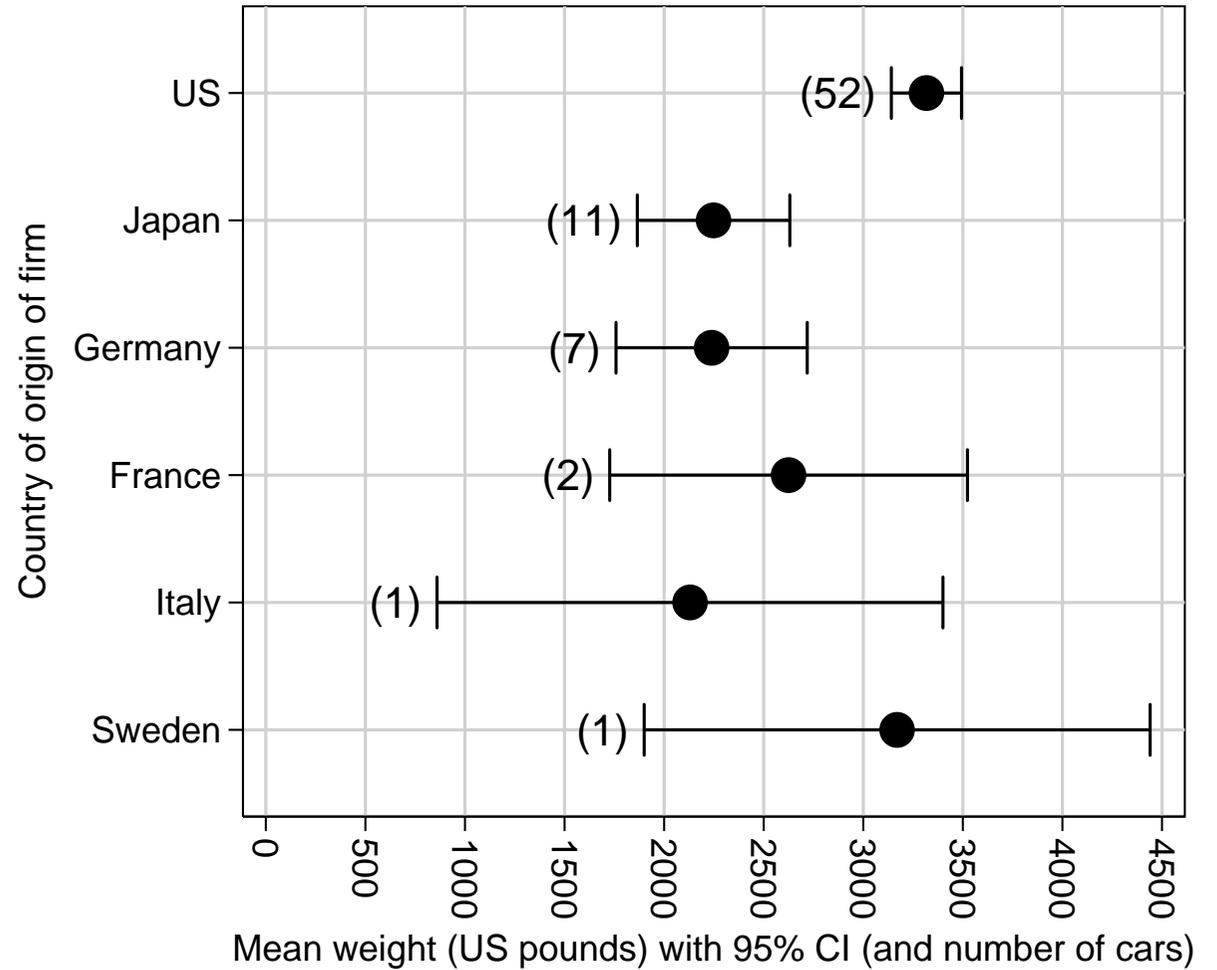
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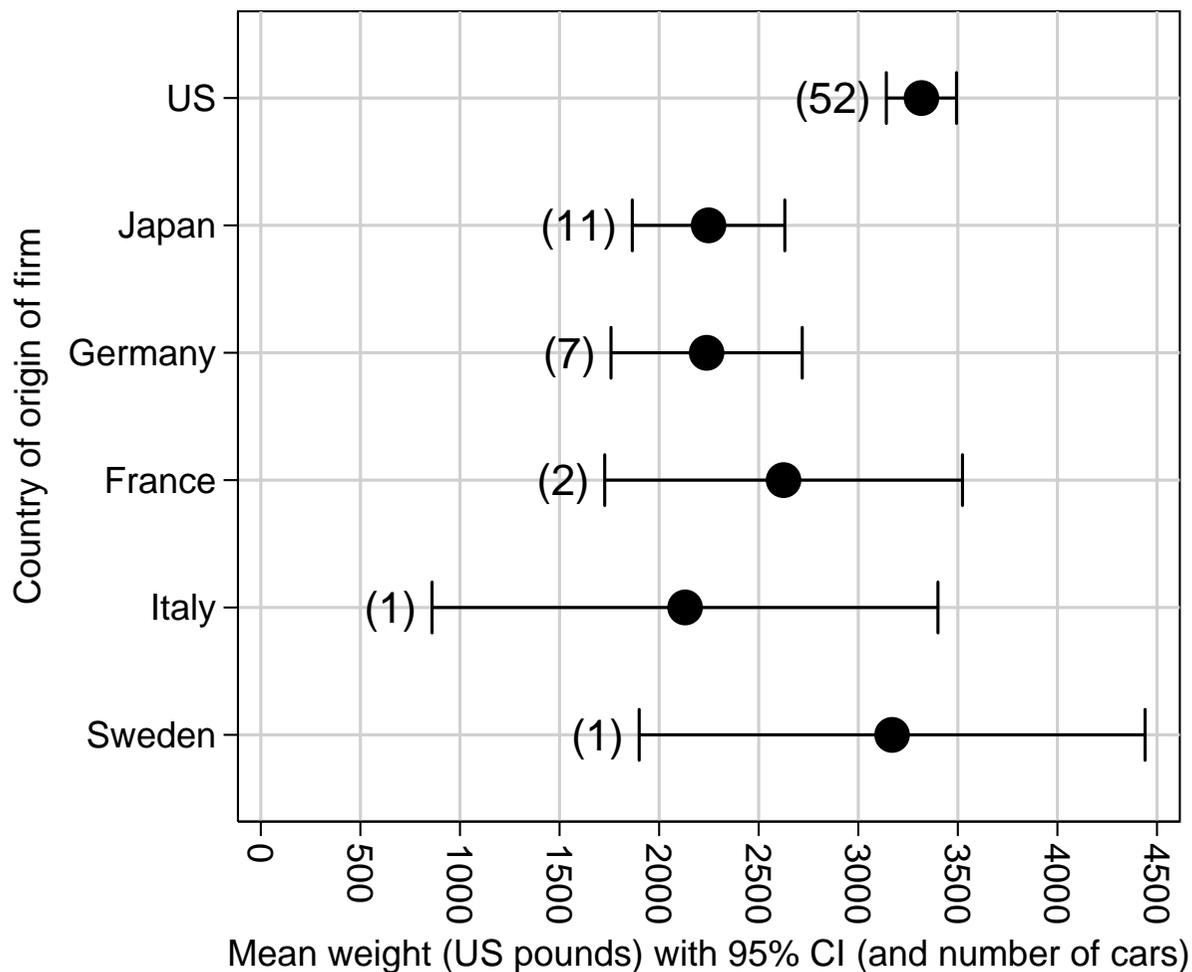
This is easier to understand than the previous output. *However ...*

## A resultsplot of the resultsset



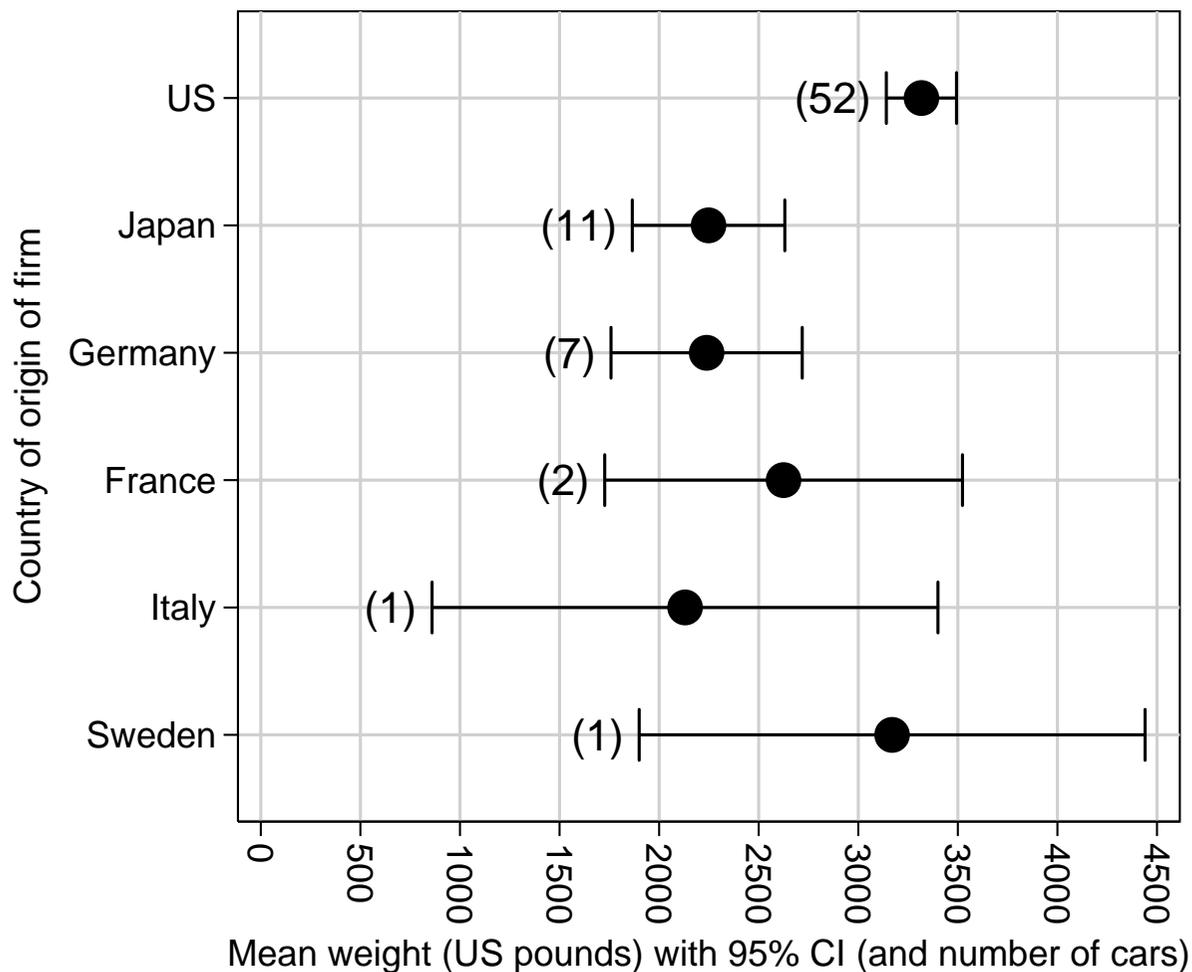
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- ... this plot was produced from the resultsset, using `ecplot` (downloadable from SSC).



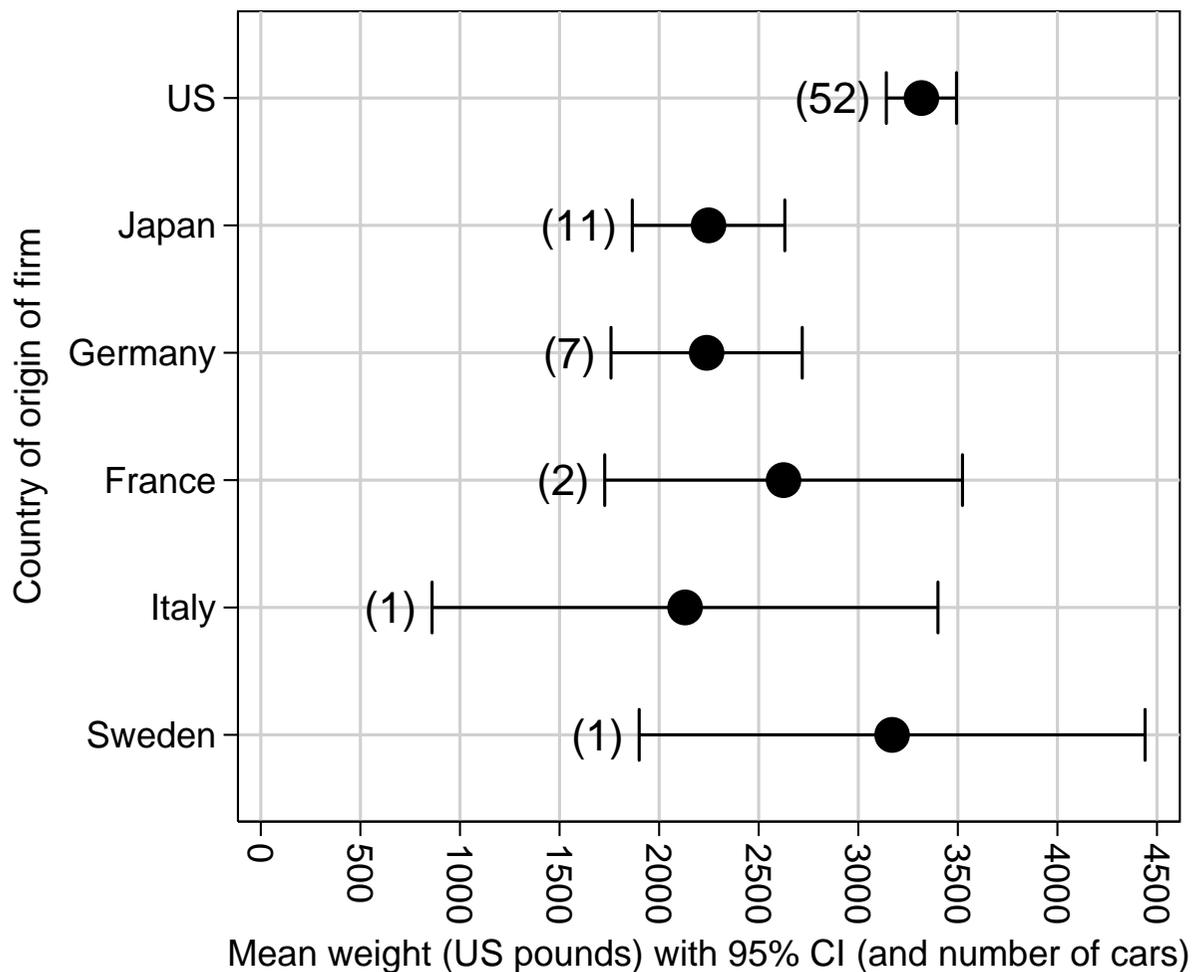
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- `eclplot` requires an input dataset with one observation per confidence interval and data on `_estimates` and `_confidence limits`.



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- `eclplot` requires an input dataset with one observation per confidence interval and data on `_estimates` and `_confidence limits`.
- The frequencies were added using the `plot()` option of `eclplot`.



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We convert the confidence limits to string, adding commas and parentheses, using the SSC package `sdecode`. Then we use the SSC package `listtex` to output the resultsset to a resultsspreadsheet in the L<sup>A</sup>T<sub>E</sub>X `tabular` row style:

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```
. listtex country _freq estimate min95 max95, rstyle(tabular) type ///
> head( ///
>   "\begin{tabular}{rrrrr}" ///
>   "\textit{Country}&\textit{N}&\textit{Mean}&\textit{(95\%)}&\textit{CI)}\\" ///
> ) ///
> foot("\end{tabular}")
\begin{tabular}{rrrrr}
\textit{Country}&\textit{N}&\textit{Mean}&\textit{(95\%)}&\textit{CI)}\\
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This is not very easy to understand. *However ...*

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<i>Country</i>	<i>N</i>	<i>Mean</i>	<i>(95%</i>	<i>CI)</i>
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`listtex` (and also another SSC package `estout`) can also produce resultsspreadsheets in other generic text formats, such as plain T<sub>E</sub>X, HTML, or tab-separated (for easy conversion to Microsoft Excel spreadsheets or Word tables).

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- A **resultsplot** is a plot, with plotted points corresponding to the observations of a resultsset (or to the rows of a resultsspreadsheet).

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- For instance, the resultsset in the previous example has one observation per *country*, and data on frequencies and confidence intervals.

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- *However*, resultssets can also be created from some types of resultsspreadsheets, using `insheet`.
- A resultsset may be listed to the Stata log and/or saved to a disk file and/or written to the memory, replacing any existing dataset.
- *However*, resultssets exist mainly as a means of creating resultsplots and/or resultsspreadsheets.

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## **A resultsset created by the SSC package xcontract**

In the `auto` data with the added variable `country`, we use `xcontract` to create a resultsset with one observation per country, and data on frequencies and percentages, which is listed to the Stata log and saved to a temporary file:

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```
. tempfile tf1

. xcontract country, saving('tf1') list(, clean noobs)
```

Listing of results:

country	_freq	_percent
US	52	70.27
Japan	11	14.86
Germany	7	9.46
France	2	2.70
Italy	1	1.35
Sweden	1	1.35

```
file C:\DOCUME~1\rnewson\LOCALS~1\Temp\ST_000000m1.tmp saved
```

**The SSC package parmest**

## The SSC package `parmest`

The `parmest` package is used with estimation commands such as `regress`. It creates a resultsset with one observation per model parameter, and data on estimates, confidence intervals and  $p$ -values. We can start by running the regression model as before:

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We then use `parmest` to create a resultsset with one observation per country, and data on mean weights and confidence limits. This is listed to the Stata log, and also written to the memory, replacing the existing data:

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```
. parmest, label norestore format(estimate min95 max95 %8.2f) ///  
> list(parm label estimate min95 max95, clean noobs)
```

parm	label	estimate	min95	max95	
c_1	country==	1.0000	3317.12	3140.98	3493.25
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Note the variable `label`, which contains, for each observation, the variable label of the  $X$ -variable of the parameter for that observation. The  $X$ -variables here are indicators (dummies) for values of `country` (from 1 to 6).

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- And *many* users want to plot confidence intervals against predictor variables!

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- The SSC package `factext` can reconstruct categorical factors such as `country` from the `label` variable in a `parmest` resultsset.

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- The SSC package `descsave` is an extended version of `describe`, which can describe a list of variables and write a do-file.
- This do-file, if run in another dataset, will reconstruct the storage types, display formats, value labels and variable labels for any variables of the same names and modes (numeric or string) which may exist in that dataset.
- The SSC package `factext` can reconstruct categorical factors such as `country` from the `label` variable in a `parmest` resultsset.
- It can then run a do-file created by `descsave` to reconstruct the storage types, display formats, value labels and variable labels.

**Running descsave in the extended auto data**

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In the extended auto dataset, we run `descsave` on the added variable `country`, writing the do-file to a temporary file:

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```
. tempfile df1
```

```
. descsave country, do('df1')
```

variable name	storage type	display format	value label	variable label
country	byte	%8.0g	country	Country of origin of firm

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```

variable name	storage type	display format	value label	variable label
country	byte	%8.0g	country	Country of origin of firm

`descsave` describes the storage type, display format, value label and variable label of `country`, and creates a temporary do-file 'df1' to reconstruct these attributes in another dataset.

**The do-file created by descsave**

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We type the temporary do-file:

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```
. type 'df1'  
cap la de country 1 "US", modify  
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cap la de country 3 "Germany", modify  
cap la de country 4 "France", modify  
cap la de country 5 "Italy", modify  
cap la de country 6 "Sweden", modify  
cap recast byte country  
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```

If this do-file is run in another dataset, and a numeric variable `country` exists in that dataset, then that numeric variable `country` will have the storage type, display format, value labels and variable label of the variable `country` in the auto data.

## Reconstructing country in the parmest resultsset

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In the `parmest` resultsset, we run `factext` to reconstruct country, from the `label` variable, and list the resultsset:

## Reconstructing country in the parmest resultsset

In the parmest resultsset, we run `factext` to reconstruct country, from the label variable, and list the resultsset:

```
. factext, do('df1')
```

```
. list parm label country estimate min95 max95, clean noobs
```

parm	label	country	estimate	min95	max95
c_1	country==	US	3317.12	3140.98	3493.25
c_2	country==	Japan	2248.18	1865.22	2631.14
c_3	country==	Germany	2238.57	1758.50	2718.64
c_4	country==	France	2625.00	1726.88	3523.12
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The variable `country` has been reconstructed from the variable `label`, with its variable labels. So we can now identify the countries of the confidence intervals.

## Match–merging resultssets

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We now merge the `parmest` resultsset with the `xcontract` resultsset that we saw earlier, and add the frequencies to the confidence intervals:

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```
. sort country

. merge country using 'tf1'
(label country already defined)

. sort country

. list country _freq estimate min95 max95, clean noobs
```

country	_freq	estimate	min95	max95
US	52	3317.12	3140.98	3493.25
Japan	11	2248.18	1865.22	2631.14
Germany	7	2238.57	1758.50	2718.64
France	2	2625.00	1726.88	3523.12
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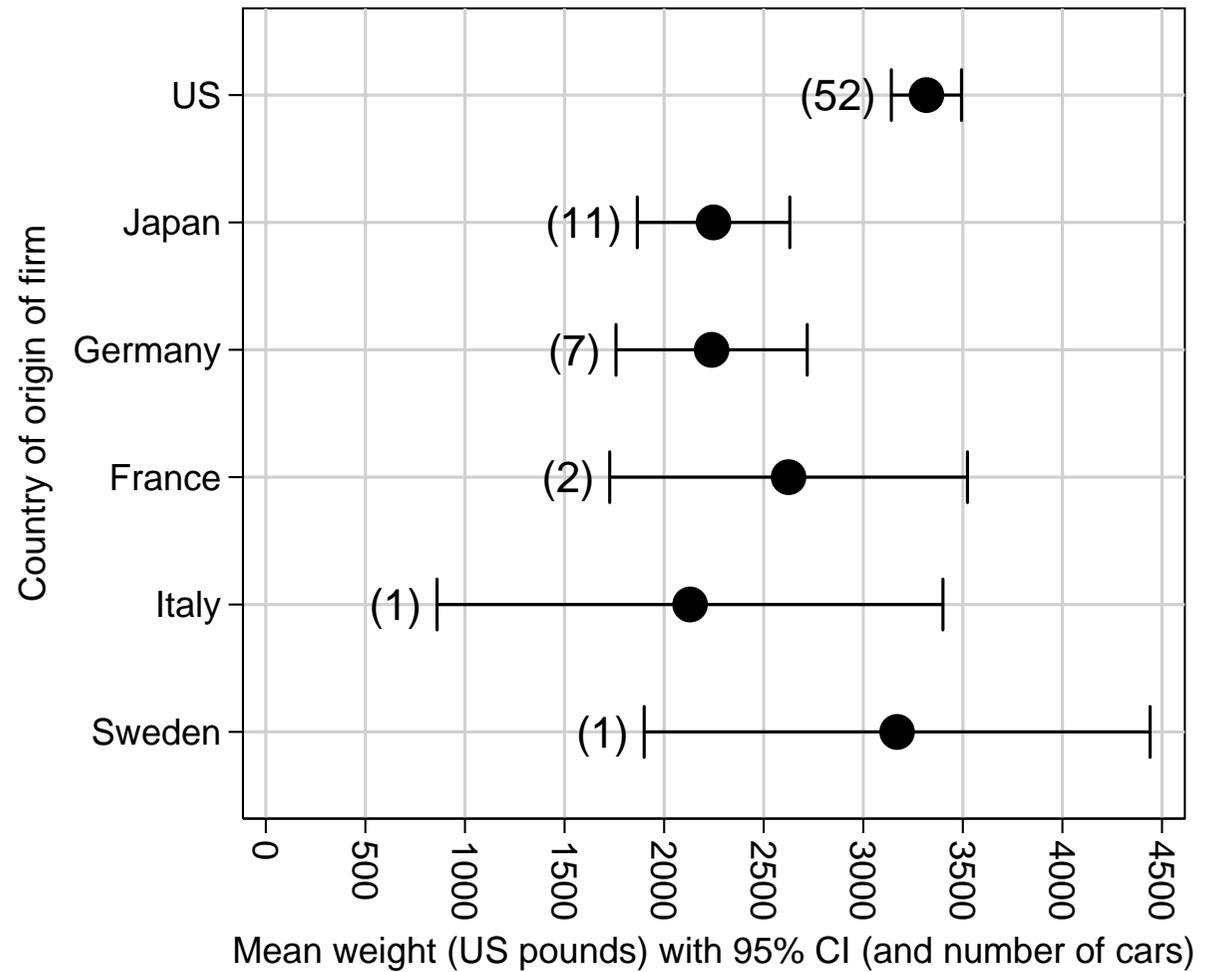
. sort country

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country	_freq	estimate	min95	max95
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Japan	11	2248.18	1865.22	2631.14
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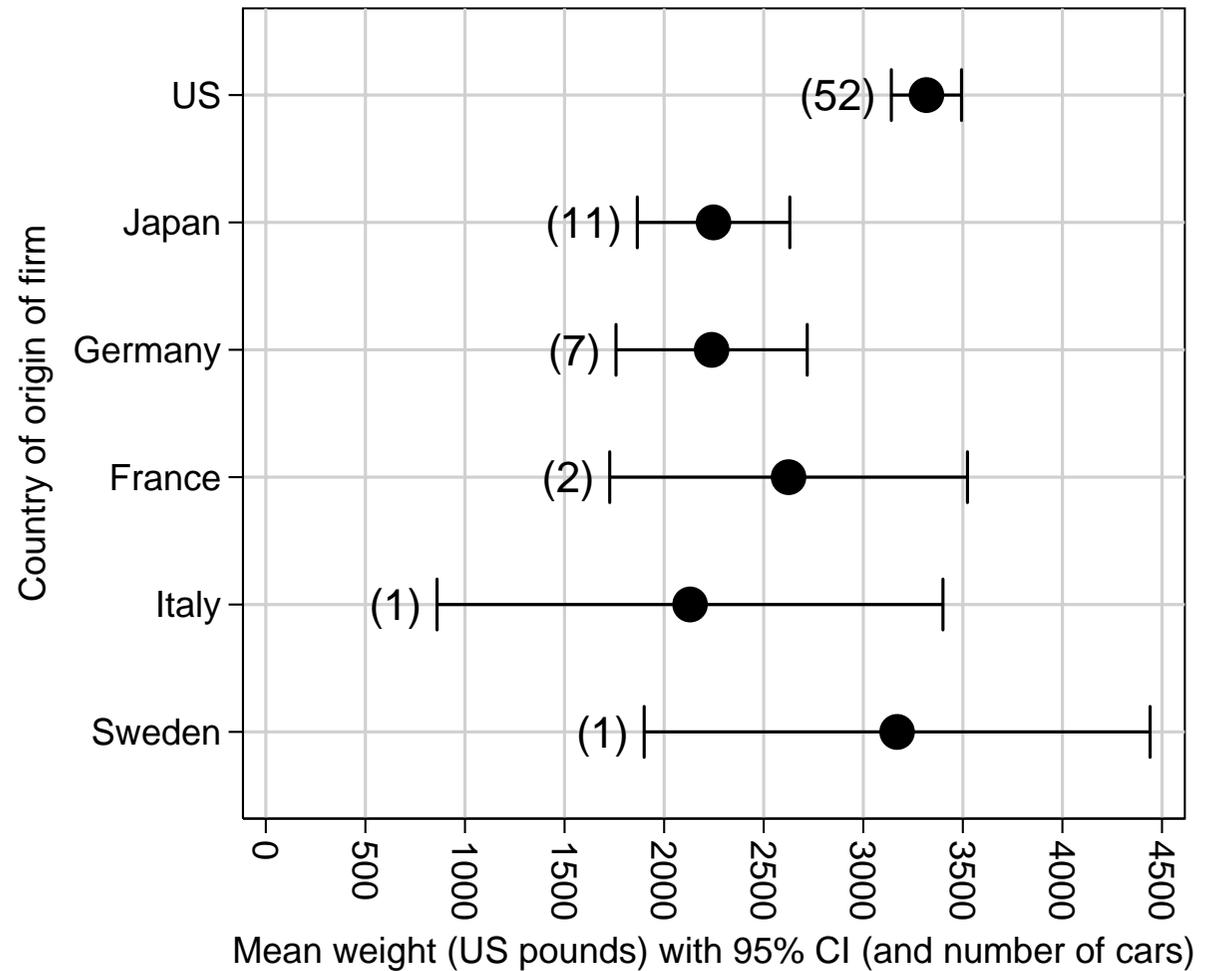
We now have the resultsset which created the resultsplot that we saw earlier.

## The power of resultssets



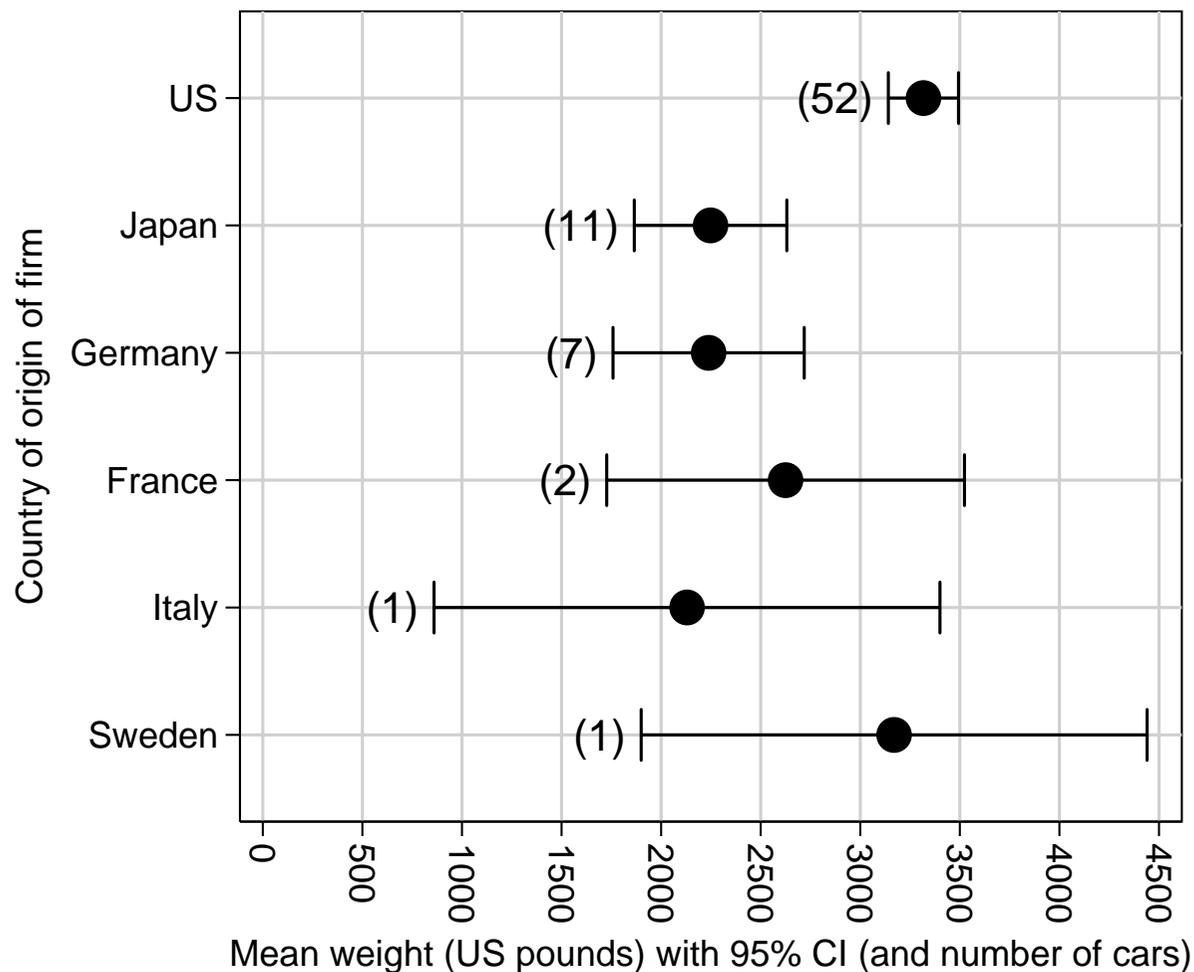
## The power of resultssets

- We used `parmest` to produce a resultsset of confidence intervals.



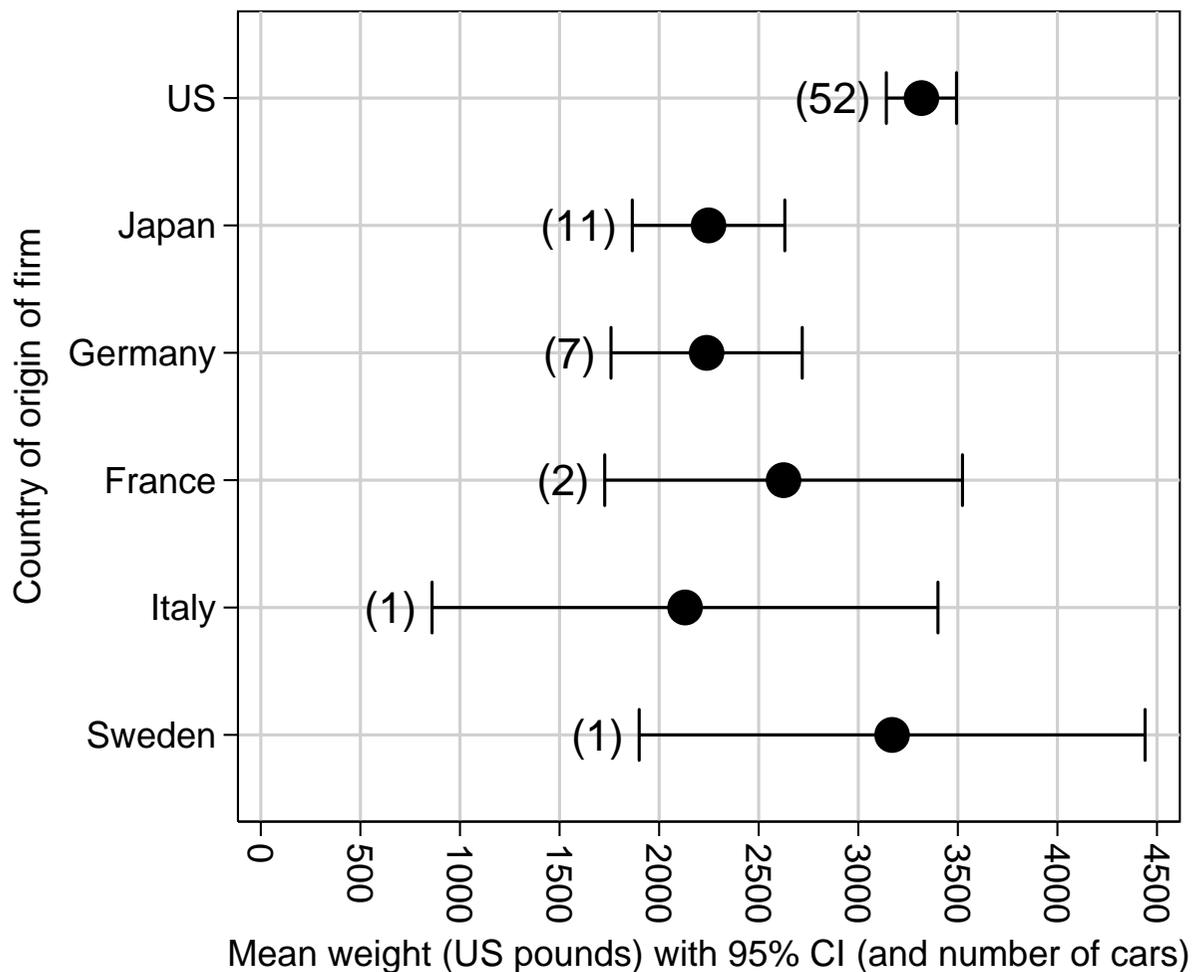
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- We used `parmest` to produce a resultsset of confidence intervals.
- We reconstructed the vertical-axis variable `country`, using `descsave` and `factext`.



## The power of resultssets

- We used `parmest` to produce a resultsset of confidence intervals.
- We reconstructed the vertical-axis variable `country`, using `descsave` and `factext`.
- We then added the frequencies by merging in a resultsset produced by `xcontract`.



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- *However*, resultsspreadsheets can also be created from resultssets (as well as *vice versa*), using `outsheet` for tab-delimited and comma-delimited formats, or the SSC package `listtex` for other formats (eg `TEX` and HTML).

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- Resultsspreadsheets can be copied (or linked) into `TEX` or HTML documents, or converted into Microsoft Excel spreadsheets or Word tables.

## Common table formats (or row styles) for resultsspreadsheets

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Each of these table row styles has a row–begin string, a column–delimiter string and a row–end string, given here as Stata expressions. The row–begin and row–end strings may be empty strings.

<i>Row style</i>	<i>Row–begin</i>	<i>Column–delimiter</i>	<i>Row–end</i>
Tab–delimited	" "	char(9)	" "
HTML table	"<tr><td>"	"</td><td>"	"</td></tr>"
L <sup>A</sup> T <sub>E</sub> X tabular	" "	"&"	"\\ "
Plain T <sub>E</sub> X halign	" "	"&"	"\cr"

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Note that resultsspreadsheets of all styles are generic text tables. Therefore, any text can appear between the row–begin, column–delimiter and row–end strings, except for row–begin, column–delimiter and row–end strings.

## The SSC package estout

## **The SSC package estout**

The `estout` package (Jann, 2005a; Jann, 2005b) is very powerful indeed. It produces a very wide range of resultsspreadsheets from the results of any estimation command. We will demonstrate a very simple example. First, we run the regression command that we used earlier:

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```
. regress weight c_*, noconst nohead
```

weight	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
c_1	3317.115	88.26823	37.58	0.000	3140.979 3493.252
c_2	2248.182	191.9154	11.71	0.000	1865.221 2631.143
c_3	2238.571	240.5786	9.30	0.000	1758.504 2718.639
c_4	2625	450.0814	5.83	0.000	1726.877 3523.123
c_5	2130	636.5112	3.35	0.001	859.8616 3400.138
c_6	3170	636.5112	4.98	0.000	1899.862 4440.138

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```
. estout using estout1.txt, ///  
> cells("b(fmt(%8.2f)) ci_l(fmt(%8.2f)) ci_u(fmt(%8.2f))") ///  
> replace label collabels(, lhs(label)) mlabels(, none)
```

label	b	min95	max95
country==US	3317.12	3140.98	3493.25
country==Japan	2248.18	1865.22	2631.14
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The resultsspreadsheet is output to the Stata log, and also to a disk file `estout1.txt`, where it can be accessed and edited by a spreadsheet package (possibly even Microsoft Excel). *However ...*

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```
. insheet using estout1.txt, clear  
(4 vars, 6 obs)
```

```
. list, clean noobs
```

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country==Japan	2248.18	1865.22	2631.14
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Or, alternatively, the user may edit `estout1.txt` manually in a spreadsheet package, and *then* convert it to a resultsset. We can then produce ...

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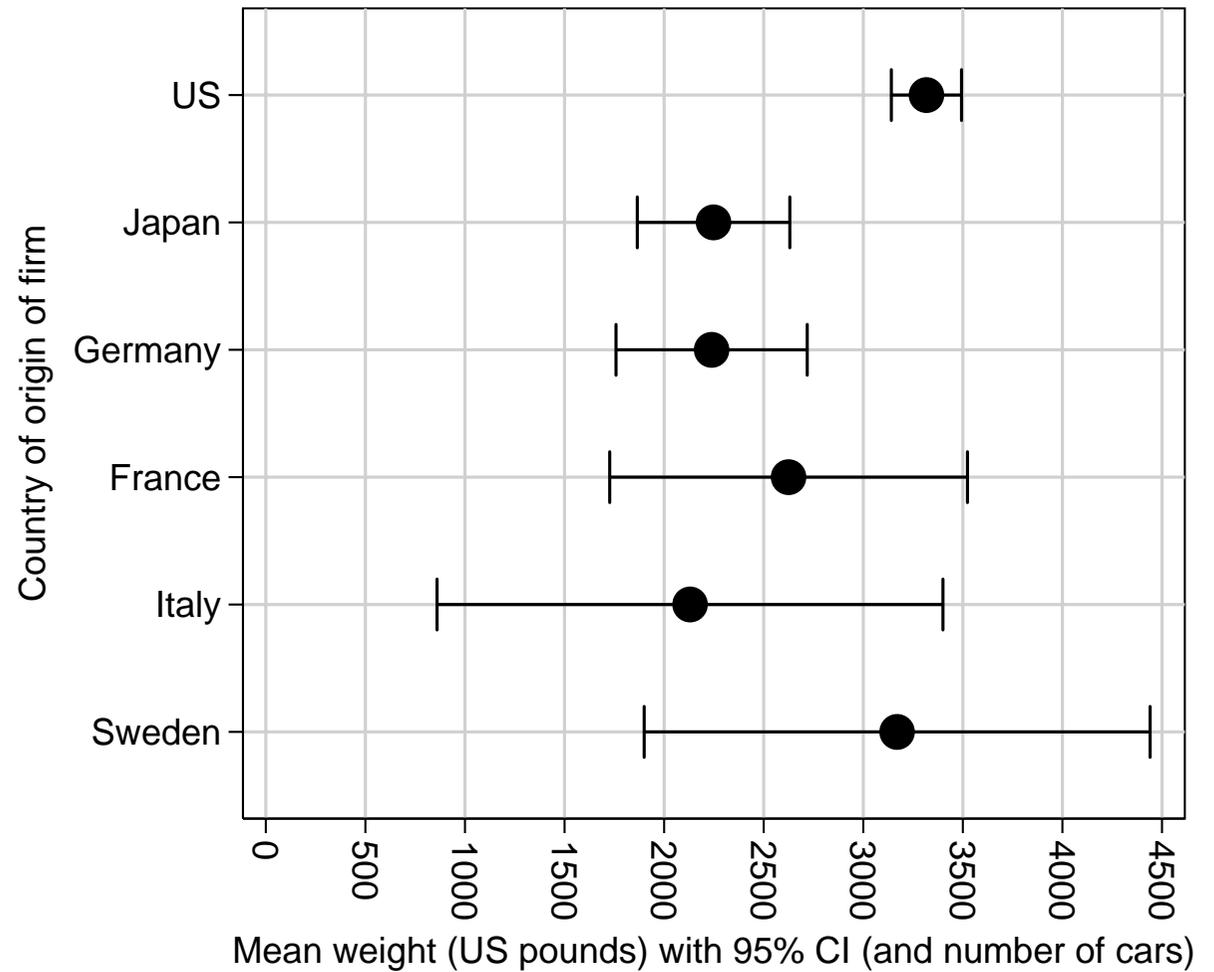
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## Resultsplots

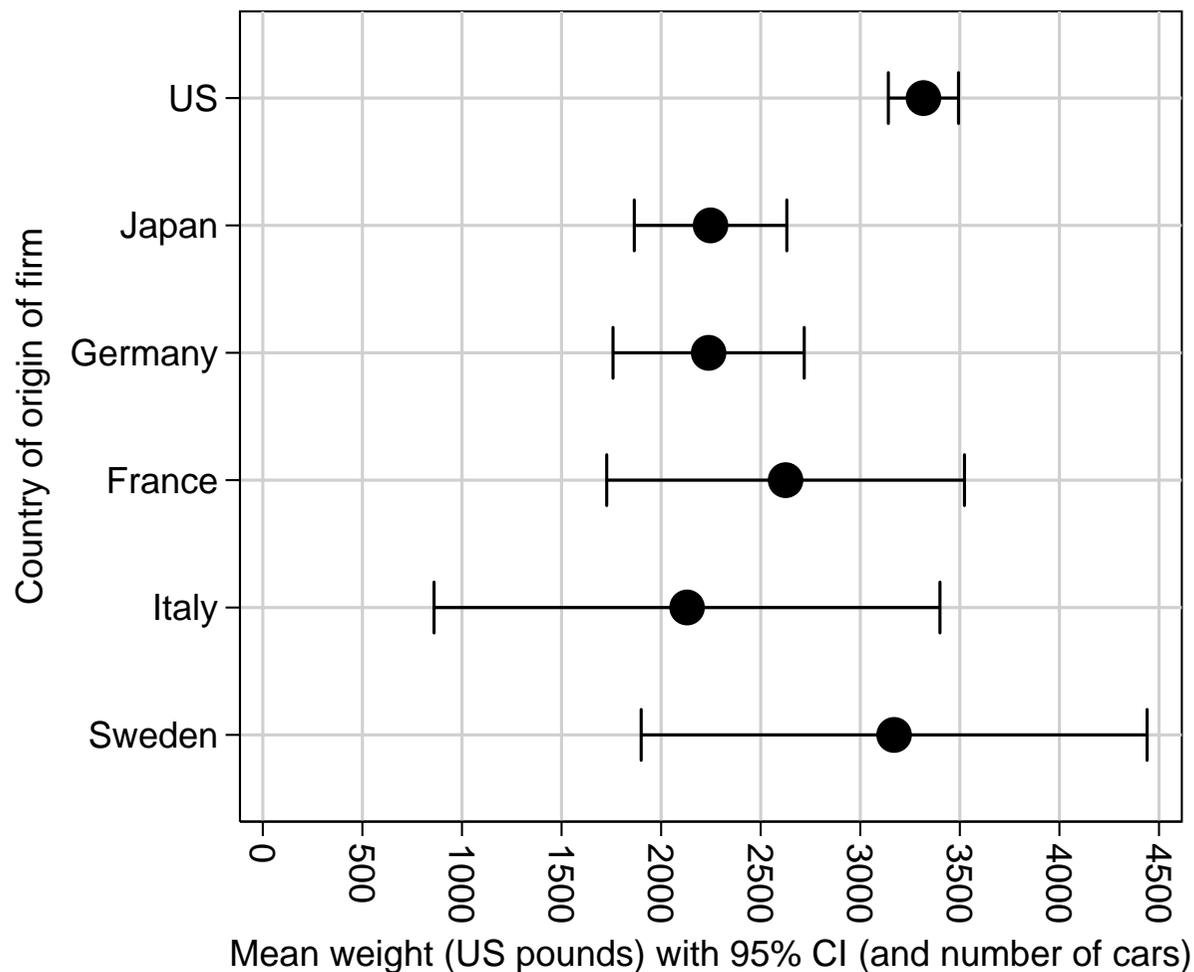
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- In Stata (at present), a resultsplot can be produced *only* from a resultsset (and *not* directly from a resultsspreadsheet).
- *However*, as we have seen, it is easy to produce a resultsset from a resultsspreadsheet.
- This makes life much easier for Stata users who do not like to do a lot of programming.

## A resultsplot of the resultsset from the resultsspreadsheet



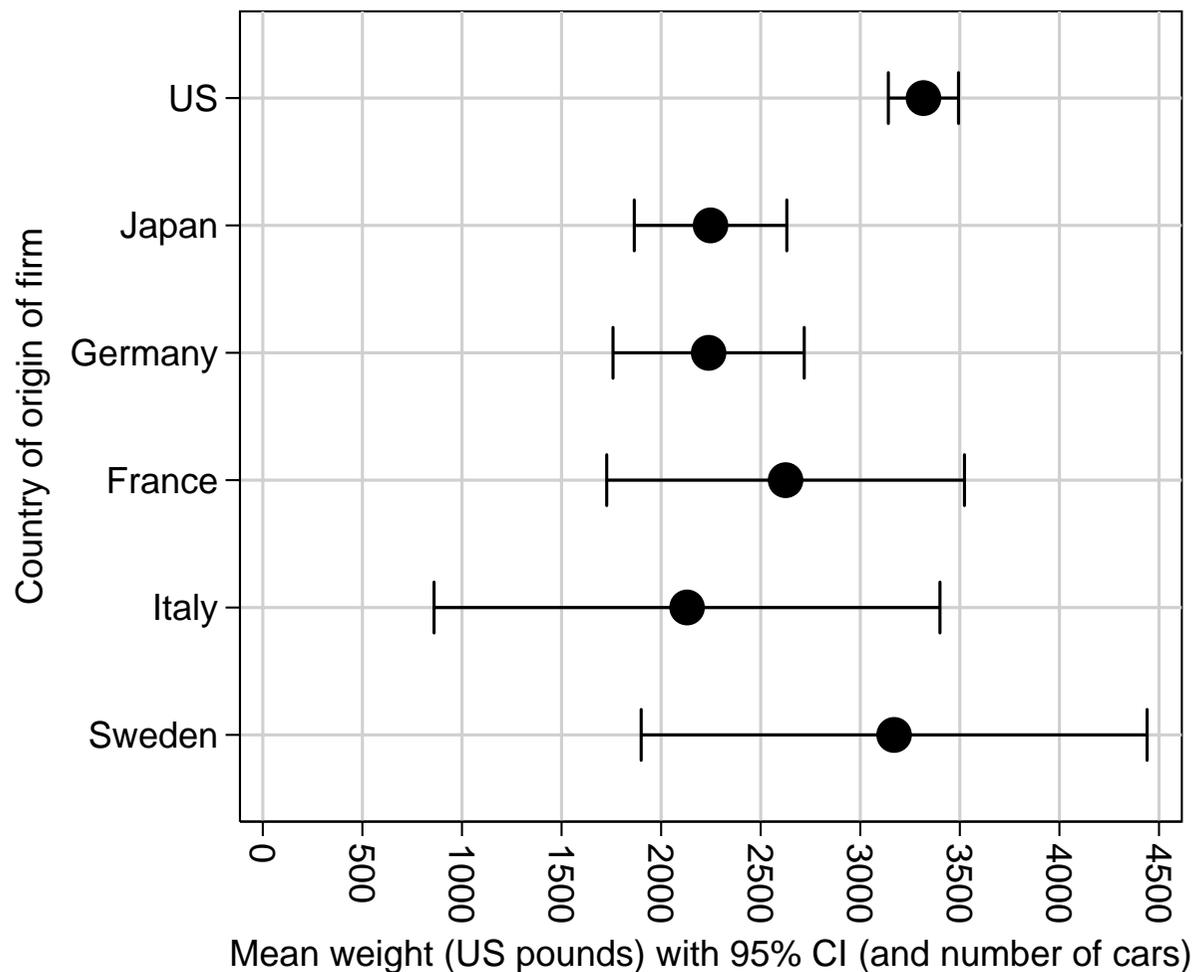
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- This plot was produced by `ec1plot` from the resultsset generated from the `estout` resultsspreadsheet.



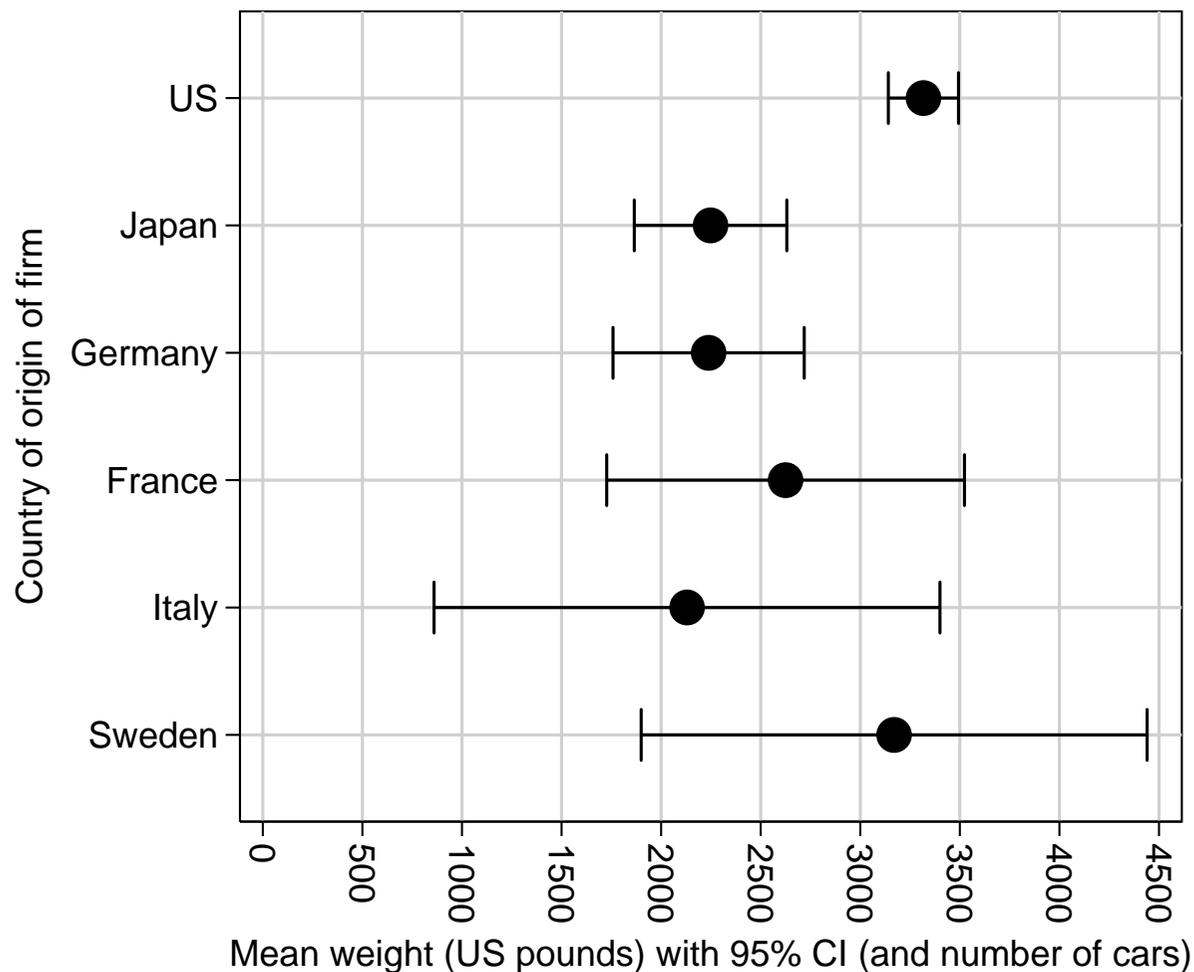
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- (And these could have been added using Ben Jann's SSC package `estadd`.)



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This is a destructive process, and is usually programmed in a do-file between a `preserve` and a `restore`. *However*, it produces the range of tables that I want to produce.

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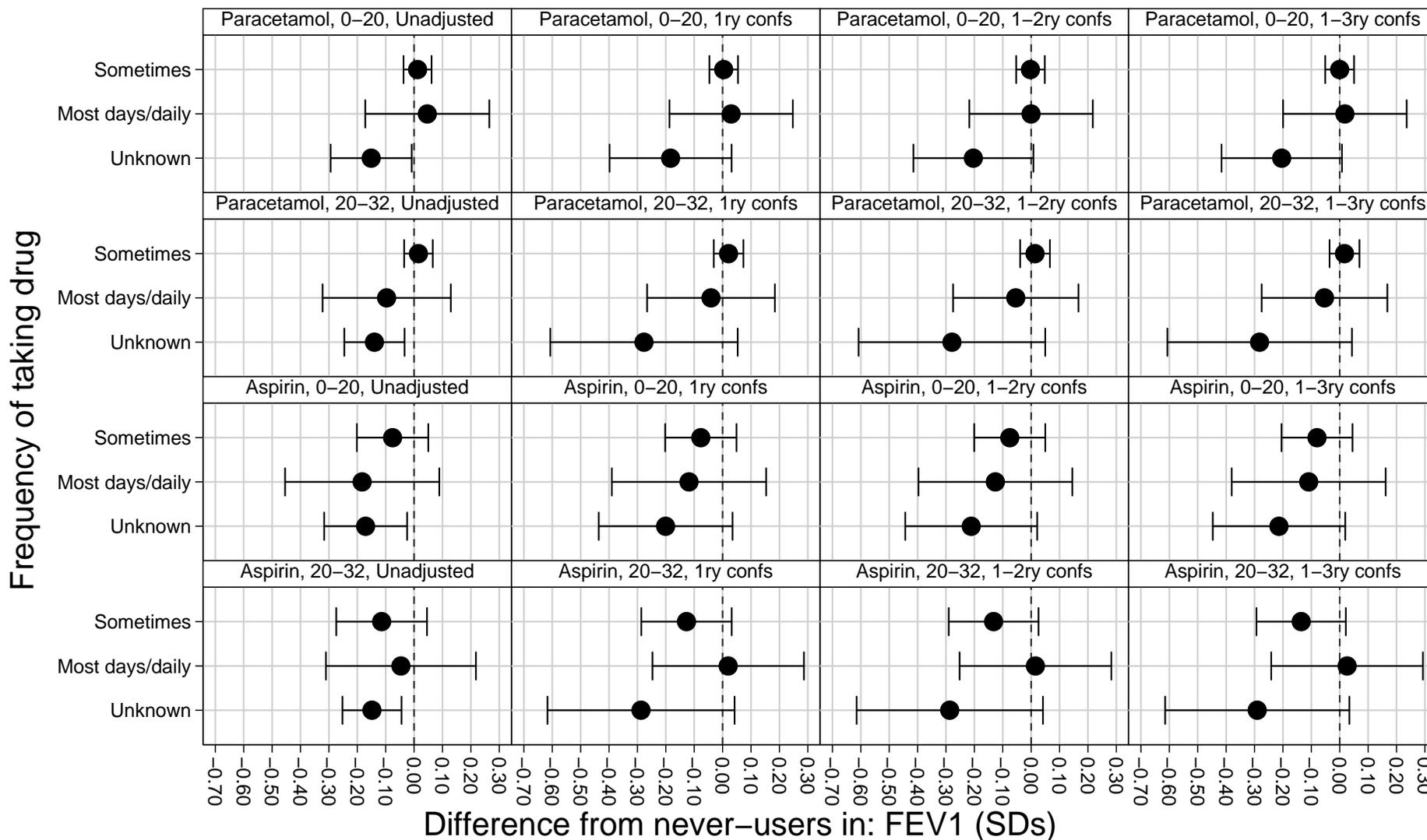
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- These differences could be unadjusted, *or* adjusted for one of 3 non–empty nested sets of confounders.
- We therefore calculated confidence intervals for  $2 \times 2 \times 3 \times 4 = 48$  differences, generated by 2 analgesics, 2 stages of pregnancy, 3 non–reference exposure levels, and 4 confounder sets.

## A resultsplot with $2 \times 2 \times 3 \times 4 = 48$ confidence intervals



Graphs by drug, Gestation week, and Confounder set

metaparm – confidence intervals for linear combinations of  
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- The SSC package `metaparm` is complementary to `lincom`, and to the SSC package `lincomest` (Newson, 2003).
- It inputs a resultsset with 1 observation for each of a set of *independently-estimated* parameters, and data on their estimates and standard errors.

**metaparm – confidence intervals for linear combinations of  
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- Or even for interactions, defined as differences between differences (or ratios between ratios) between parameters.

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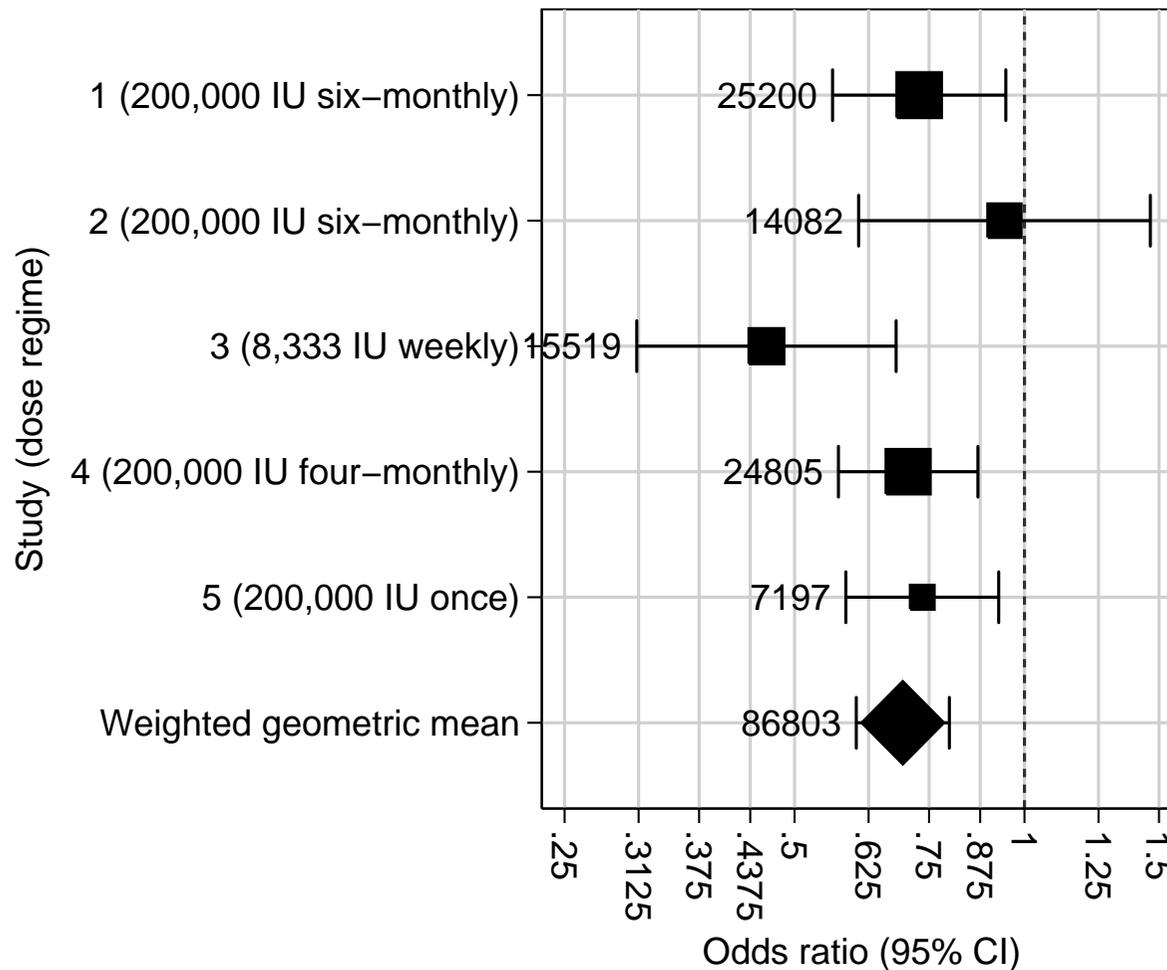
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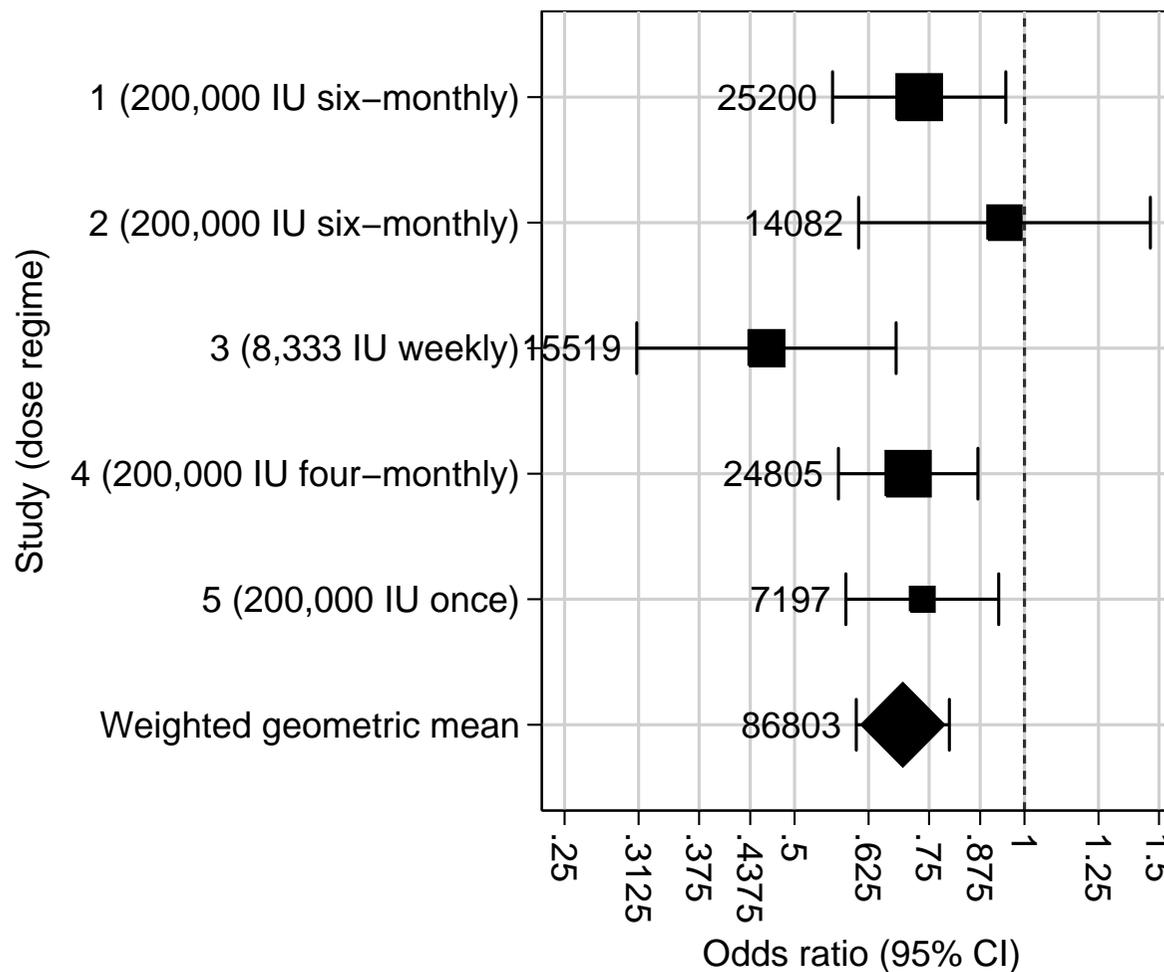
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- The data were reproduced in Bland (2000).
- We meta-analysed the data using `parmby`, `metaparm` and `ec1plot`.

## A Cochrane forest plot produced using metaparm and eclplot



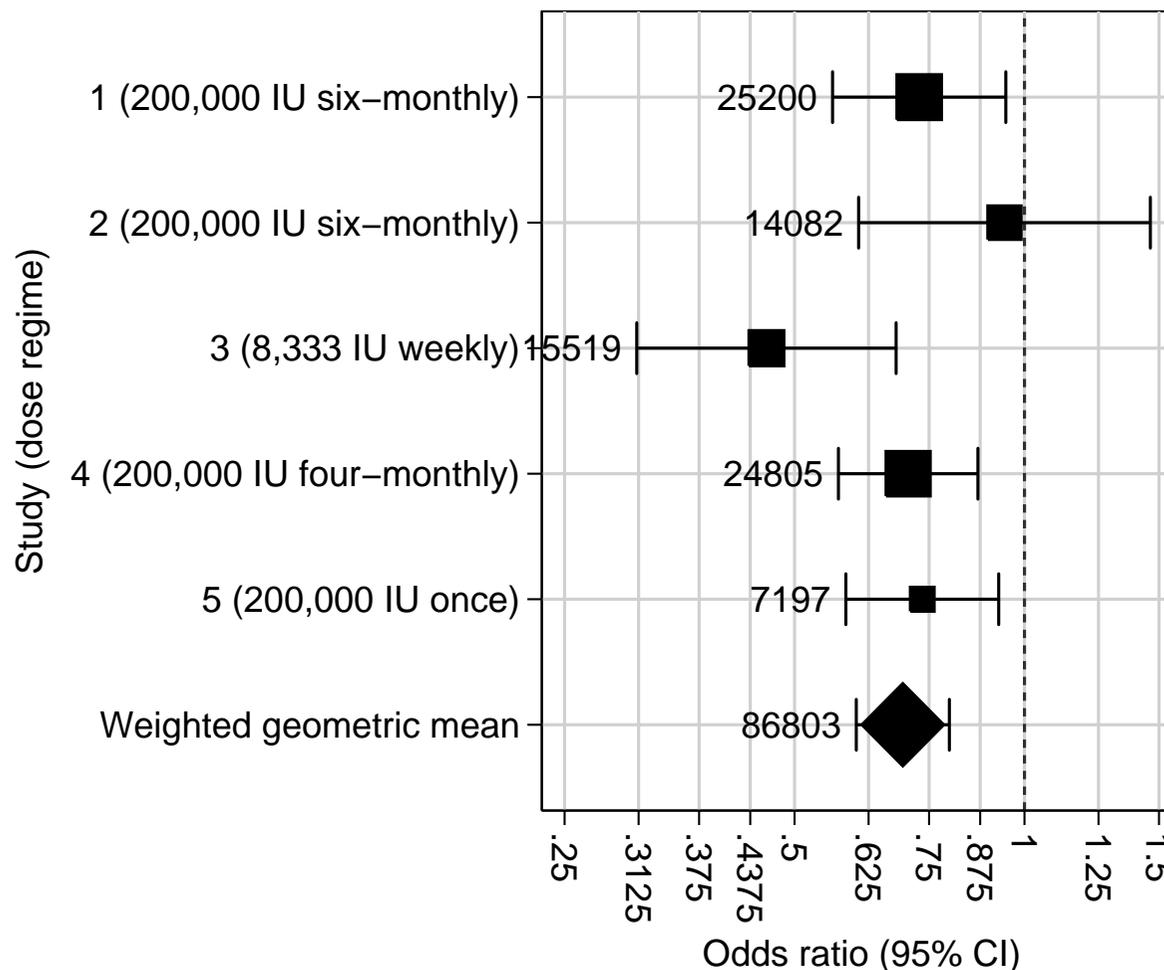
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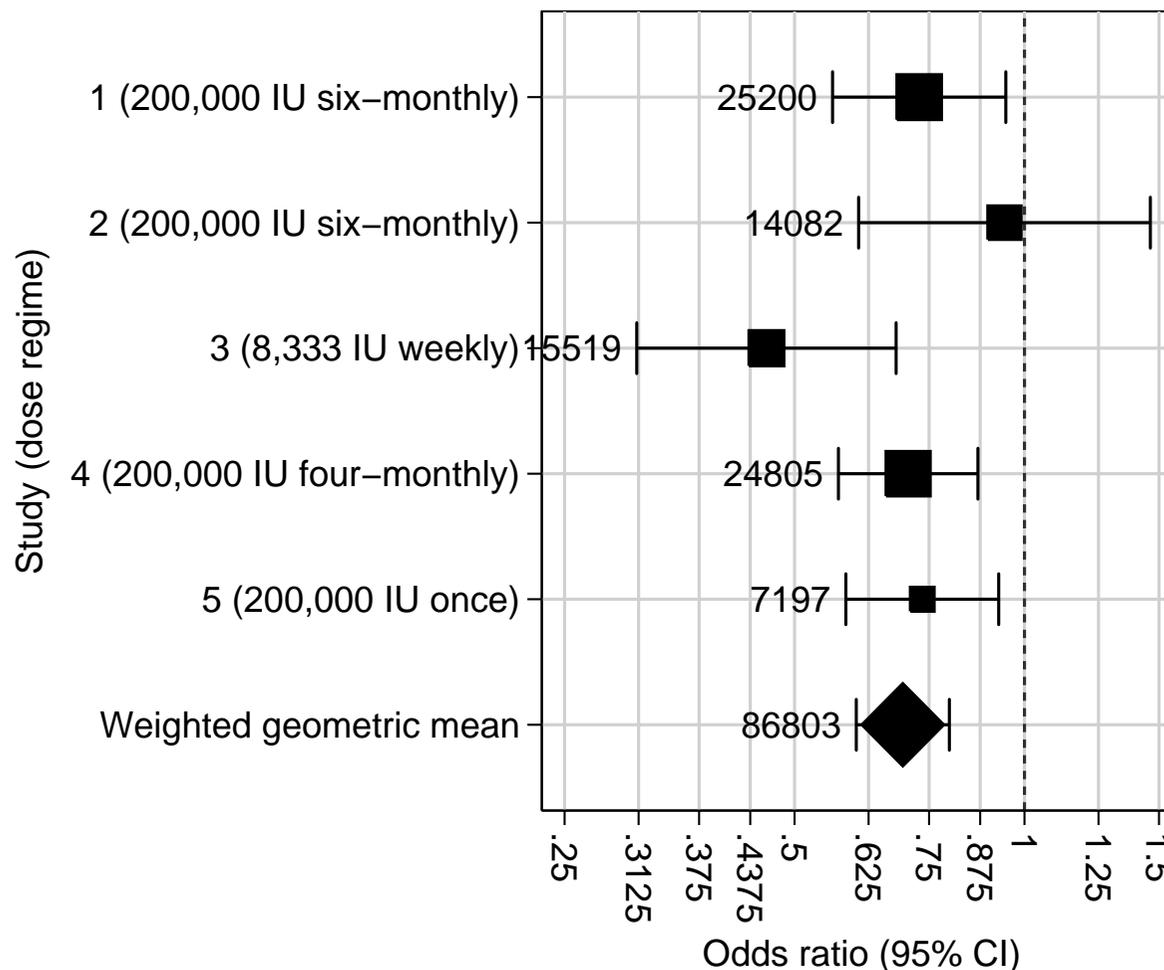
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- Note that `eclplot` can weight symbols by study size.



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- The two resultsprocessing dogmas are complementary, and each one has advantages and disadvantages.

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- For instance, variables in a resultsset have storage types, display formats, variable labels and value labels, often inherited from variables of the same names in the original dataset.
- Resultssets are most useful when plots and tables are being mass-produced, as in the child cohort example.

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## Conclusions

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- The resultsset–central dogma is followed by a minority of programmers, who want to write a do–file to do *everything*, with little or no manual intervention.
- And the same packages are often useful under either dogma!

## References (1)

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All Stata User Meeting presentations referenced (and this one) can be downloaded from <http://www.stata.com/meeting/proceedings.html>

The packages `descsave`, `estadd`, `estout`, `factext`, `listtex`, `metaparm`, `parmest`, `regaxis`, `scheme_rbn1mono`, `sencode`, `sdecode`, `xcollapse` and `xcontract` (mentioned and/or used in this presentation) can be downloaded from SSC using the `ssc` command in Stata.