**Description**

`estat bubbleplot` produces bubble plots after simple meta-regression with one continuous moderator performed by using `meta regress`. The bubble plot is a scatterplot of effect sizes against a moderator of interest overlaid with the predicted regression line and confidence-interval bands. In a bubble plot, the marker sizes, “bubbles”, are proportional to study weights.

**Quick start**

Fit a random-effects meta-regression with a continuous moderator, \( x \)

```
meta regress x, random
```

Construct a bubble plot for \( x \)

```
estat bubbleplot
```

As above, but specify that the size of the marker representing studies be proportional to the random-effects weights instead of the default fixed-effects weights

```
estat bubbleplot, reweighted
```

Construct a bubble plot with a 90\% confidence interval

```
estat bubbleplot, level(90)
```

**Menu**

Statistics > Meta-analysis
Syntax

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Options

`reweighted` is used with random-effects meta-regression. It specifies that the sizes of the bubbles be proportional to the weights from the random-effects meta-regression, \( w_j^* = \frac{1}{(\hat{\sigma}_j + \hat{\tau}^2)} \). By default, the sizes are proportional to the precision of each study, \( w_j = \frac{1}{\hat{\sigma}_j^2} \).

`regline` and `noregline` display or suppress the rendition of the regression line. The default, `regline`, is to display the regression line. Option `noregline` implies option `noci`.

`ci` and `noci` display or suppress confidence intervals. The default, `ci`, is to display them.

`level(#)` specifies the confidence level, as a percentage, for confidence intervals. The default is as declared for the meta-analysis session; see `Declaring a confidence level for meta-analysis` in [META] `meta data`. Also see option `level()` in [META] `meta set`.

`n(#)` specifies the number of points at which to evaluate the CIs. The default is n(100).

`lineopts(line_options)` affects the rendition of the plotted regression line; see [G-3] `line_options`.

`ciopts(ci opts)` affects the rendition of the CI band in the bubble plot. `ci opts` are any options as defined in [G-2] `graph twoway rline` and option `recast(rarea)` as described in [G-3] `advanced_options`. 
Add plots

`addplot(plot)` allows adding more graph `twoway` plots to the graph; see `[G-3] addplot_option`. Y axis, X axis, Titles, Legend, Overall

`twoway_options` are any of the options documented in `[G-3] twoway_options`, excluding `by()`. These include options for titling the graph (see `[G-3] title_options`) and for saving the graph to disk (see `[G-3] saving_option`).

Remarks and examples

Remarks are presented under the following headings:

Introduction
Examples of using `estat bubbleplot`

Introduction

A bubble plot (Berkey et al. 1995; Thompson and Sharp 1999; Thompson and Higgins 2002) is used after simple meta-regression with a continuous moderator to describe the relation between the effect size and the corresponding moderator. It is used as a tool to assess how well the regression model fits the data and to potentially identify influential and outlying studies. The bubble plot is a scatterplot with the study-specific effect sizes plotted on the `y` axis and the moderator of interest from the meta-regression plotted on the `x` axis. The sizes of the markers or “bubbles” are proportional to the precision of each study. The more precise (larger) the study, the larger the size of the bubble. The predicted regression line and confidence bands are overlaid with the scatterplot.

`estat bubbleplot` produces bubble plots after simple meta-regression with a continuous moderator performed by using `meta regress`. Traditionally, the weights used to determine the sizes of the bubbles are the inverses of the effect-size variances, $1/\sigma_j^2$. After a random-effects meta-regression, you can specify the `reweighted` option to instead use the random-effects weights, $1/(\hat{\sigma}_j^2 + \tau^2)$.

The predicted regression line and the 95% confidence intervals are plotted by default. You can specify the `level()` option to obtain other confidence intervals. You can control the look of the lines by specifying the options `lineopts()` and `ciopts()`. You can also suppress the lines by specifying the options `noregline` and `noci`.

Examples of using `estat bubbleplot`

In the examples that follow, we demonstrate how to create and customize bubble plots after a meta-regression. Consider the BCG dataset from Examples of using `meta regress` in [META] `meta regress`.

```
. use https://www.stata-press.com/data/r16/bcgset
  (Efficacy of BCG vaccine against tuberculosis; set with -meta esize-)
  . meta query, short
  -> meta esize npost nnegt nposc nnegc, esize(lnrratio) studylabel(studylbl)

  Effect-size label: Log Risk-Ratio
  Effect-size type: lnrratio
  Effect size: _meta_es
  Std. Err.: _meta_se
  Model: Random-effects
  Method: REML
```
To create these plots, we first fit the random-effects meta-regression shown in example 1 of \texttt{[META] meta regress}, but our focus here is not on the interpretation of these plots but on the variety of bubble plots that can be created.

\begin{verbatim}
.meta regress latitude_c
  Effect-size label: Log Risk-Ratio
  Effect size: _meta_es
  Std. Err.: _meta_se

Random-effects meta-regression
  Method: REML
  Number of obs = 13
  Residual heterogeneity:
    tau2 = .07635
    I2 (%) = 68.39
    H2 = 3.16
    R-squared (%) = 75.63
    Wald chi2(1) = 16.36
    Prob > chi2 = 0.0001

|                | Coef.   | Std. Err. | z    | P>|z| | [95% Conf. Interval] |
|----------------|---------|-----------|------|------|----------------------|
| latitude_c     | -.0291017 | .0071953 | -4.04 | 0.000 | -.0432043 -.0149991 |
| _cons          | -0.7223204 | .1076535 | -6.71 | 0.000 | -.9333174 -.5113234 |

Test of residual homogeneity: Q_res = chi2(11) = 30.73 Prob > Q_res = 0.0012
\end{verbatim}

\textbf{Example 1: A basic bubble plot}

To construct a bubble plot after performing a simple meta-regression, we simply type

\begin{verbatim}
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  Effect-size label: Log Risk-Ratio
  Effect size: _meta_es
  Std. Err.: _meta_se

Random-effects meta-regression
  Method: REML
  Number of obs = 13
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Test of residual homogeneity: Q_res = chi2(11) = 30.73 Prob > Q_res = 0.0012
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The graph shows the log risk-ratios plotted against the mean-centered latitudes of the studies’ locations. By default, the regression line and corresponding confidence intervals are plotted. We could suppress these and plot just the bubbles with options noregline and noci, respectively. The regression line provides a good fit of the data because most studies are relatively close to it. The log risk-ratios for the BCG vaccine decline with increasing latitude. For more interpretation of the above bubble plot, refer to example 1 of \texttt{[META] meta regress postestimation}. 

\begin{verbatim}
Example 1: A basic bubble plot

To construct a bubble plot after performing a simple meta-regression, we simply type

\texttt{. estat bubbleplot}

\begin{figure}
  \centering
  \includegraphics{bubble_plot.png}
  \caption{Bubble plot showing log risk-ratios against mean-centered latitudes.}
\end{figure}

The graph shows the log risk-ratios plotted against the mean-centered latitudes of the studies’ locations. By default, the regression line and corresponding confidence intervals are plotted. We could suppress these and plot just the bubbles with options noregline and noci, respectively. The regression line provides a good fit of the data because most studies are relatively close to it. The log risk-ratios for the BCG vaccine decline with increasing latitude. For more interpretation of the above bubble plot, refer to example 1 of \texttt{[META] meta regress postestimation}.
\end{verbatim}
Example 2: Reweighting the bubbles

By default, the bubble sizes are proportional to trial precisions, $1/\hat{\sigma}_j^2$. With the `reweighted` option, we can make the bubble sizes proportional to the weights from the random-effects meta-regression, $1/(\hat{\sigma}_j^2 + \hat{\tau}^2)$. For example, continuing with example 1, we can reweight the bubbles as follows:

```
    . estat bubbleplot, reweighted
```

With random-effects weights, the sizes of the bubbles are more uniform across the studies as compared with precision (fixed-effects) weights used in example 1. This will always be true except when $\hat{\tau}^2 = 0$, in which case the bubble sizes will be identical with both types of weights.
Example 3: Using addplot() to add labels for the trials

Below, we discuss how you can add labels to the trials, which are represented by the hollow circles on a bubble plot. Typically, we use `mlabel(varname)` to add marker labels. For example, if we wish to label the trials according to their trial ID, `trial`, we type

```
estat bubbleplot, mlabel(trial)
```

Specifying the `mlabel()` option causes all the markers to have the same size. One way to get around this is by using the `addplot()` option.

We can use `addplot()` to overlay an exact copy of the properly weighted bubble plot but without plotting the markers and symbols, that is, using the `msymbol(none)` option. We can then add labels to these nonplotted symbols to obtain the desired plot. Here is our minimal `addplot()` specification:

```
addplot(scatter _meta_es lattitude_c, msymbol(none) mlabel(trial))
```

The full specification is

```
local opts msymbol(none) mlabel(trial) mlabcolor(navy) legend(order(1 2 3))
'estat bubbleplot, addplot(scatter _meta_es latitude_c, `opts' mlabpos(2))
```
We used additional options to fine-tune the bubble plot. The `mlabcolor(navy)` option controls the color of the study labels (navy here). The `legend(order(1 2 3))` option prevents the display of a legend key for the added scatterplot in the legend box at the bottom of the plot. Finally, the `mlabpos(2)` option specifies that marker labels be drawn at the 2 o’clock position.

Example 4: Adjusting label positions

Continuing with example 3, let’s customize the labels further. For example, marker labels 3 and 10 (and 6 and 1) are not easily distinguishable. You may provide individual marker label positions for each study by using the `mlabvpos(varname)` option. `varname` must be created to hold the individual positions (an integer number between 0 to 12) of each marker.

Let’s generate our position variable.

```
. generate byte pos = 2
. quietly replace pos = 9 in 10/13
. quietly replace pos = 6 if inlist(trial,1,2,5)
```

We generated a new variable `pos` to hold the individual positions of each marker label. We chose to draw labels at 9 o’clock for trials 10 to 13, at 6 o’clock for trials 1, 2, and 5, and at 2 o’clock for the other trials.

We now use a similar specification of `addplot()` from example 3, but here we add `mlabvpos(pos)` and `mlabgap(*2)` to double the gap size between the marker labels and the markers so that the trial labels do not touch the hollow circles; see trials 6 and 8.

```
. estat bubbleplot, addplot(scatter _meta_es latitude_c, mlabvpos(pos) > mlabgap(*2) 'opts')
```

![Bubble plot](image)
We can modify other aspects of the graph such as the legend. Let’s place the legend inside the plot region. We also specify if `inlist(trial,7,12,13)` with `estat bubbleplot` to display trial labels only for specific trials (for example, outliers, trials with large weights, and so on).

```
.local legopts legend(ring(0) position(2) cols(1) size(small) symxsize(*0.3))
estat bubbleplot, addplot(scatter _meta_es latitude_c if inlist(trial,7,12,13), mlabvpos(pos) mlabgap(*2) 'opts') 'legopts'
```

Within the `legend()` option (see [G-3] `legend_options`), `ring(0)` and `position(2)` specify that the legend be placed inside the plot region at the 2 o’clock position. `cols(1)` specifies that all legend keys be displayed in one column. `size(small)` specifies that a small font be used for the legend key text, and `symxsize(*0.3)` sets the width of the key symbols to 30% of their default width.

### Methods and formulas

`estat bubbleplot` produces a scatterplot with the effect sizes, \( \hat{\theta}_j \), stored in the system variable `_meta_es` on the y axis and a moderator from the meta-regression on the x axis. By default, the bubble size is proportional to \( w_j = 1/\hat{\sigma}_j^2 \). For a random-effects meta-regression, if you specify the `reweighted` option, the weights \( w_j^* = 1/(\hat{\sigma}_j^2 + \hat{\tau}^2) \) will be used.

For a simple meta-regression with moderator \( x_1 \), the plotted predicted line is \( \hat{\beta}_0 + \hat{\beta}_1 x_1 = x_j \hat{\beta} \). The CIs are computed as

\[
x_j \hat{\beta} \pm z_{1-\alpha/2} \hat{SE}(x_j \hat{\beta})
\]

where the computation of \( \hat{SE}(x_j \hat{\beta}) \) is described in [META] `meta regress postestimation`. The n() option specifies how many evaluation points are used to construct the CI plots. By default, 100 points are used. When the se() or tdistribution option is specified with `meta regress`, the confidence intervals use the \( t_{K-2,1-\alpha/2} \) critical value.
References


Also see

[META] meta regress — Meta-analysis regression
[META] meta regress postestimation — Postestimation tools for meta regress
[META] meta — Introduction to meta
[META] Glossary
[META] Intro — Introduction to meta-analysis