

# How can I obtain Somers' D after performing a logistic regression in Stata?

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## RECOMMENDED CITATION

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To obtain Somers' D after performing a logistic regression in Stata, one can use the "somersd" command. This command takes the predicted probability and the outcome variable from the logistic regression as input and calculates Somers' D, which is a measure of rank correlation between the two variables. It also provides a confidence interval and a p-value for the calculated D value. This allows for an assessment of the strength and significance of the relationship between the predicted probability and the outcome variable in the logistic regression model. The "somersd" command is a useful tool for evaluating the predictive ability of logistic regression models in Stata.

## How can I get a Somers' D after logistic regression in Stata? | Stata FAQ

Recently we received the following question: The output from SAS proc logistic gives a values for Somers' D. How can I get this using Stata?

The Somers' D, in logistic regression, provides an estimate of the rank correlation of the observed binary response variable and the predicted probabilities. Thus, it can be used as an indicator of model fit.

Its not difficult to get a Somers' D in Stata once you download the user contributed program somersd written by Roger Newson. To get the program just type, `ssc install somersd`, in Stata's command window and follow the prompts to download the program. Alternatively, you can type `search somersd` and follow the prompts (see How can I use the search command to

**search for programs and get additional help? for more information about using search).**

**Once you have downloaded somersd, run your logistic regression, compute the predicted probability and then run somersd with your binary response variable and the predicted probability.**

### **Example**

**We will use the hsb2 dataset for this example. It doesn't have a good binary response variable so we will create one by dichotomizing write at the value 60 and calling it honors (for honors English). Then we will use it in a logistic regression with read, female and prog (the type of program each student is in) as predictors. After the logit command we will use predict to get the predicted probabilities.**

**use <https://stats.idre.ucla.edu/stat/data/hsbdemo>, clear**

**logit honors read female i.prog**

**Iteration 0: log likelihood = -115.64441**

**Iteration 1: log likelihood = -86.845312**

**Iteration 2: log likelihood = -84.560995**

**Iteration 3: log likelihood = -84.542357**

**Iteration 4: log likelihood = -84.542348**

**Iteration 5: log likelihood = -84.542348**

**Logistic regression Number of obs = 200**

**LR chi2(4) = 62.20**

**Prob > chi2 = 0.0000**

**Log likelihood = -84.542348 Pseudo R2 = 0.2689**

-----  
**honors | Coef. Std. Err. z P>|z|**  
 -----+-----

**read | .1352861 .0242218 5.59 0.000 .0878123 .18276**

**female | 1.08343 .4094357 2.65 0.008 .2809511 1.885909**

|

**prog |**

**2 | .5559416 .5053125 1.10 0.271 -.4344527 1.546336**

**3 | .0016408 .6611702 0.00 0.998 -1.294229 1.29751**

|

**\_cons | -9.41691 1.481922 -6.35 0.000 -12.32142  
 -6.512397**  
 -----

**predict pprob**

Now that we have the predicted probabilities, pprob, we can run somersd.

```
somersd honors pprob
```

Somers' D with variable: honors

Transformation: Untransformed

Valid observations: 200

Symmetric 95% CI

| Jackknife

```
honors | Coef. Std. Err. z P>|z|
```

```
-----+-----
pprob | .6761648 .05778 11.70 0.000 .5629182 .7894115
-----+-----
```

The value of Somers' D is 0.676. We can compare this value of Somers' D to one from a model that uses only prog as a predictor.

```
logit honors i.prog
```

Iteration 0: log likelihood = -115.64441

Iteration 1: log likelihood = -107.7993

**Iteration 2: log likelihood = -107.57279**

**Iteration 3: log likelihood = -107.5719**

**Iteration 4: log likelihood = -107.5719**

**Logistic regression Number of obs = 200**

**LR chi2(2) = 16.15**

**Prob > chi2 = 0.0003**

**Log likelihood = -107.5719 Pseudo R2 = 0.0698**

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**honors | Coef. Std. Err. z P>|z|**  
 -----+-----

**prog |**

**2 | 1.206168 .4577746 2.63 0.008 .3089465 2.10339**

**3 | -.3007541 .5988045 -0.50 0.615 -1.474389 .8728812**

**|**

**\_cons | -1.691676 .4113064 -4.11 0.000 -2.497822 -  
 .8855303**  
 -----

**predict pprob2**

**(option pr assumed; Pr(honors))**

**somersd honors pprob2**

**Somers' D with variable: honors**

**Transformation: Untransformed**

**Valid observations: 200**

**Symmetric 95% CI**

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**| Jackknife**

**honors | Coef. Std. Err. z P>|z|**

-----+-----  
**pprob2 | .3228084 .0737763 4.38 0.000 .1782095 .4674073**

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**As you can see the Somers' D for this model is much smaller than for the previous one.**