

# Fundamentals of Statistical Modeling (VT20)

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## Lab 3 (Extra material on flexible modeling with splines)

Load the dataset and the `mlci` command

{{1}}

Assume that  $f(\text{age})$  follows a generalized extreme values distribution. Estimate the parameters  $\mu$  and  $\sigma$ . Constrain  $\sigma$  to be positive. Inflate the probability of death during the first year of life, while constraining it to be between 0 and 1.

{{2}}

Generate the estimated density and the transform  $u1 = \hat{F}(y)$  (we'll use it to assess the goodness-of-fit).

Remember: we're assuming that, for  $\text{age} \geq 1$ , the variable  $\text{age}$  is Standard-Exponential-distributed after we apply the transform  $G(y)$ . The CDF of a Standard Exponential is  $F(y) = 1 - \exp(-y)$

{{3}}

Now, let's include a spline transformation of  $\text{age}$  with 4 degrees of freedom and let's see whether this improves the fit of our generalized extreme values model. Jointly test the 3 parameters  $\eta_1, \eta_2, \eta_3$  to assess whether adding the 3 RCS transforms improves the fit of this model with respect to the "base" model (see above).

We need to help Stata a little by providing reasonable initial values for the model's parameters.

{{4}}

Generate the estimated density and the transform  $u2 = \hat{F}(y)$ .

{{5}}

Plot the 2 estimated densities over the sample histogram and the quantile plot.

{{6}}