Arithmetic expressions in Biogeme

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Series on Biogeme

The package Biogeme (`biogeme.epfl.ch`) is designed to estimate the parameters of various models using maximum likelihood estimation. It is particularly designed for discrete choice models.

This document describes how Biogeme handles arithmetic expressions and deals with potential numerical issues. The concepts have been implemented in `cythonbiogeme 1.0.2`, used by `biogeme 3.2.13`.

1 Introduction

The core of the Biogeme software package is the calculation of formulas for each observation in a database. In estimation mode, the formula is the log likelihood function. And its derivatives are necessary for the optimization algorithm as well as the calculation of useful statistics. In simulation mode, the formulas are any indicator that the analyst deeems useful to calculate (choice probabilities, elasticities, etc.) We refer the reader to Bierlaire (2018) and Bierlaire (2023) for more details about the use of Biogeme for model estimation and the calculation of indicators.

To allow the user to use Biogeme on a wide variety of model specifications, the formulas are composed of elementary arithmetic operations. These building blocks are organized in a complex tree structure, where each of them receives inputs from others, generates output, that is forwarded to the next layer. For instance, the formula

$$-x + \frac{\exp(y-1)}{2}$$

can be represented as illustrated in Figure 1.

Each node of the formula is associated with a specific simple operation, and is in charge of calculating its value, and its derivatives.

Computers are working with finite arithmetic. It means that computers have limitations in the way they represent and operate on numbers due to their finite hardware resources and the design of numerical representations. Therefore, the actual implementation of the arithmetic operations are not necessarily an exact duplicate of their mathematical equivalent, that consider a continuous space of real numbers, that can take any value.

The objective of this document is to describe how each arithmetic expression is handled by Biogeme.

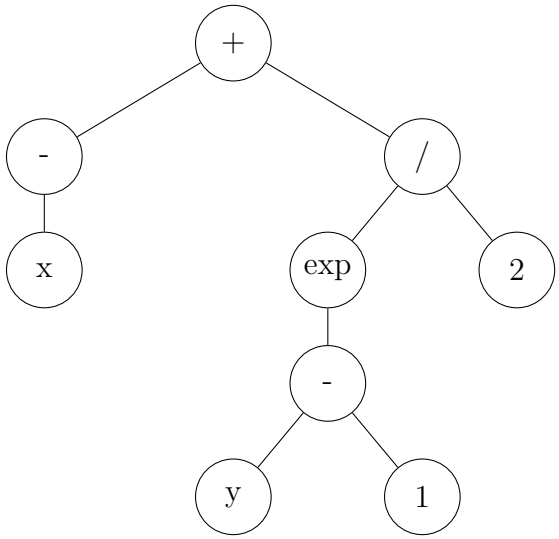

Figure 1: Tree representation of a formula

2 Numerical limits

The computer representation of a real number is called a "floating-point" representation. It is divided into three parts. The values of the parameters below correspond to a 64-bit representation:

- A sign bit s, that indicates the sign of the number (0 for positive, 1 for negative).
- An exponent e covering $k = 11$ bits, that represents the exponent of the number in a biased form. By bias, it is meant that negative and positive powers of two are possible. The bias b is added to obtain a positive number. The bias is $b = 2^{k-1} - 1 = 1023$, so that the exponents range between -1022 to +1023.
- A mantissa (m), using $p - 1$ bits, where p is the precision.

So, for a 64-bit representation, $s + k + p - 1 = 64$, so that $p = 53$. Therefore, the value of a floating point number is

$$(-1)^s (1 + m) 2^{e-b}.$$

This representation allows for only a finite quantity of real numbers to be represented: $2^{64} \approx 10^{19}$ numbers. And this imposes numerical limits. In Python, it is possible to retrieve information about those limits using `numpy`. If you type `print(np.finfo(float))`, you obtain:

```
Machine parameters for float64
---------------------------------------------------------------
precision =  15   resolution = 1.0000000000000001e-15
machep =    -52   eps =        2.2204460492503131e-16
negep =     -53   epsneg =     1.1102230246251565e-16
minexp =  -1022   tiny =       2.2250738585072014e-308
maxexp =   1024   max =        1.7976931348623157e+308
Nexp =       11   min =        -max
smallest_normal = 2.2250738585072014e-308   smallest_subnormal
   = 4.9406564584124654e-324
---------------------------------------------------------------
```

In this document, we consider two of those limits. We denote $\mathfrak{u}$ the largest value that can be represented, that is

$$\mathfrak{u} \approx 10^{308}.$$

We also consider ε, the "machine epsilon", that is the difference between 1.0 and the next smallest representable float larger than 1.0. It is an important value, because it means that, for each $x < \varepsilon$, adding x to 1 will provide 1 as a result:

$$1 + x = 1.$$

This may clearly lead to unpredictable behavior of numerical calculations. As mentioned in the output of `numpy`, the value of the machine epsilon in 64-bit representation is

$$\varepsilon \approx 10^{-16}.$$

There is an empirical way to calculate this value, using the following script:

```
epsilon = 1
while 1.0 + epsilon != 1.0:
    epsilon /= 2.0
epsilon *= 2.0
```

3 Expressions

Arithmetic expressions in Biogeme are based on the following principles.

- A *valid* value is a value $-\sqrt{\mathfrak{u}} \leq x \leq \sqrt{\mathfrak{u}}$, that is a value between -10^{154} and 10^{154}.
- Each arithmetic expression takes as input one or several valid values, and returns a valid value or raises an exception.
- A value x such that $x \leq \sqrt{\varepsilon}$, that is $x \leq 10^{-8}$ is considered to be zero in continuous arithmetic.

Suppose that we have an expression α. In order to make values valid in the sense described above, we define a validation function ν as follows:

$$\nu(\alpha) = \begin{cases} \sqrt{u} & \text{if } \alpha \geq \sqrt{u}, \\ -\sqrt{u} & \text{if } \alpha \leq -\sqrt{u}, \\ \alpha & \text{otherwise.} \end{cases} \tag{1}$$

Based on those principles, we explicitly characterize how each arithmetic expression is implemented in Biogeme. Each expression in Biogeme is represented by an object of generic type `Expression`.

The document is organized by groups of expressions:

- Elementary expressions, including the numbers, the variables, the parameters, etc.
- The unary expressions, accepting one input value.
- The comparison expressions, accepting two input values, and used to compare two expressions.
- Other binary expressions, accepting two input values.
- The n-ary expressions, accepting more than two values.
- The logit expression, implementing the logit model.

3.1 Elementary expressions

The elementary expressions are the building blocks of any expression. They correspond to the leaves of the tree representation, such as the one illustrated in Figure 1.

Numeric values Numeric values are the most basic expressions. The syntax for numeric values is

```
Numeric(x)
```

where `x` is the value. In most cases, the user does not need to use this syntax, as Biogeme tries to identify them automatically. If the value x is not valid, in the sense defined above, an exception is triggered.

Variables Variables are refering to the columns of the data set:

```
Variable('name_of_the_variable')
```

This expression simply returns the value of the corresponding variable for the current row. No specific validity check is performed for the sake of computational efficiency.

Parameters Parameters must be estimated from data. Their first values is defined by the user. There are two categories of parameters. Free parameters are updated by the optimization algorithm. Fixed parameters are not. The syntax for parameters is

```
Beta('name_of_the_parameter', x_0, ell, u, fixed)
```

where `x_0` is the initial value of the parameter, `ell` is the lower bound on the parameter, `u` is the upper bound on the parameter, and `fixed` specifies if the parameter must be fixed (`fixed=1`) of free (`fixed=0`). If the value `x_0`, `ell` or `u` is not valid, in the sense defined above, an exception is triggered.

Random variable A random variable is used in the context of numerical integration. The syntax is

```
RandomVariable('name_of_the_random_variable').
```

Draws Random draws are used in the context of Monte-Carlo integration. The syntax is

```
bioDraws('name_of_the_draws', draw_type)
```

where `draw_type` is a string. It can refer to a user defined type of draws, or one of the native draws from the following list:

```
'UNIFORM', 'UNIFORM_ANTI', 'UNIFORM_HALTON2',
   'UNIFORM_HALTON3', 'UNIFORM_HALTON5', 'UNIFORM_MLHS',
   'UNIFORM_MLHS_ANTI', 'UNIFORMSYM', 'UNIFORMSYM_ANTI',
   'UNIFORMSYM_HALTON2', 'UNIFORMSYM_HALTON3',
   'UNIFORMSYM_HALTON5', 'UNIFORMSYM_MLHS',
   'UNIFORMSYM_MLHS_ANTI', 'NORMAL', 'NORMAL_ANTI',
   'NORMAL_HALTON2', 'NORMAL_HALTON3', 'NORMAL_HALTON5',
   'NORMAL_MLHS', 'NORMAL_MLHS_ANTI'
```

Biogeme calculates derivatives with respects to "literals", that is, variables and parameters. In the following, we denote by x_i and x_j the literals that are involved in the derivatives. Obviously, we have

$$\frac{\partial x_i}{\partial x_i} = 1, \ \frac{\partial x_i}{\partial x_j} = 0,$$

and

$$\frac{\partial^2 x_i}{\partial x_i^2} = \frac{\partial^2 x_i}{\partial x_i \partial x_j} = 0.$$

3.2 Unary expressions

Unary expressions take one value as input. Like any expression, they return a value and the derivatives.

Unary minus Syntax: if `alpha` is the input value, the unary minus is

```
-alpha
```

If α is the input value, it returns $f(\alpha) = -\alpha$. The syntax is simply

```
-alpha
```

As α is a valid value, so is $f(\alpha)$. The derivatives are:

$$\frac{\partial f}{\partial x_i} = -\frac{\partial \alpha}{\partial x_i}$$

and

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = -\frac{\partial \alpha}{\partial x_i \partial x_j}.$$

Exponential If `alpha` is the input value, the syntax is

```
exp(alpha)
```

Let α be the input value.

- If $\alpha \leq \ln(\sqrt{u})$, then
$$f(\alpha) = e^{\alpha}.$$
Also, the derivatives are
$$\frac{\partial f(\alpha)}{\partial x_i} = v\left(e^{\alpha}\frac{\partial \alpha}{\partial x_i}\right),$$
and
$$\frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} = v\left(e^{\alpha}(\frac{\partial \alpha}{\partial x_i}\frac{\partial \alpha}{\partial x_j} + \frac{\partial^2 \alpha}{\partial x_i \partial x_j})\right),$$
where v is the validation function (1).
- If $\alpha > \ln(\sqrt{u})$, then
$$f(\alpha) = \frac{\partial f(\alpha)}{\partial x_i} = \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} = \sqrt{u}.$$

Logarithm If `alpha` is the input value, the syntax is

```
log(alpha)
```

Let α be the input value.

- If $\sqrt{\varepsilon} \leq \alpha \leq e^{\sqrt{u}}$, we define

$$\begin{aligned} f(\alpha) &= \ln(\alpha), \\ \frac{\partial f(\alpha)}{\partial x_i} &= v\left(\frac{1}{\alpha}\frac{\partial \alpha}{\partial x_i}\right), \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= v\left(-\frac{1}{\alpha^2}\frac{\partial \alpha}{\partial x_i}\frac{\partial \alpha}{\partial x_j} + \frac{1}{\alpha}\frac{\partial^2 \alpha}{\partial x_i \partial x_j}\right). \end{aligned}$$

- If $\alpha > e^{\sqrt{u}}$, we define

$$\begin{aligned} f(\alpha) &= \sqrt{u}, \\ \frac{\partial f(\alpha)}{\partial x_i} &= \sqrt{u}, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= 0. \end{aligned}$$

- If $0 \leq \alpha < \sqrt{\varepsilon}$, we define

$$\begin{aligned} f(\alpha) &= -\sqrt{u}, \\ \frac{\partial f(\alpha)}{\partial x_i} &= \sqrt{u}, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= 0. \end{aligned}$$

Logarithm or zero This expression is the same as the logarithm, except that if the argument is zero, it returns 0. If `alpha` is the input value, the syntax is

```
logzero(alpha)
```

Let α be the input value.

- If $\sqrt{\varepsilon} \leq \alpha \leq e^{\sqrt{u}}$, we define

$$\begin{aligned} f(\alpha) &= \ln(\alpha), \\ \frac{\partial f(\alpha)}{\partial x_i} &= v\left(\frac{1}{\alpha}\frac{\partial \alpha}{\partial x_i}\right), \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= v\left(-\frac{1}{\alpha^2}\frac{\partial \alpha}{\partial x_i}\frac{\partial \alpha}{\partial x_j} + \frac{1}{\alpha}\frac{\partial^2 \alpha}{\partial x_i \partial x_j}\right). \end{aligned}$$

- If $\alpha > e^{\sqrt{u}}$, we define

$$\begin{aligned} f(\alpha) &= \sqrt{u}, \\ \frac{\partial f(\alpha)}{\partial x_i} &= \sqrt{u}, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= 0. \end{aligned}$$

- If $0 < \alpha < \sqrt{\varepsilon}$, we define

$$\begin{aligned} f(\alpha) &= -\sqrt{u}, \\ \frac{\partial f(\alpha)}{\partial x_i} &= \sqrt{u}, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= 0. \end{aligned}$$

- If $\alpha = 0$, we define

$$\begin{aligned} f(\alpha) &= 0, \\ \frac{\partial f(\alpha)}{\partial x_i} &= 0, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= 0. \end{aligned}$$

Sinus If `alpha` is the input value, the syntax is

```
sin(alpha)
```

Let α be the input value. We define

$$\begin{aligned} f(\alpha) &= \sin(\alpha), \\ \frac{\partial f(\alpha)}{\partial x_i} &= \cos(\alpha)\frac{\partial \alpha}{\partial x_i}, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= -\sin(\alpha)\frac{\partial \alpha}{\partial x_i}\frac{\partial \alpha}{\partial x_j} + \cos(\alpha)\frac{\partial^2 \alpha}{\partial x_i \partial x_j}. \end{aligned}$$

Cosinus If `alpha` is the input value, the syntax is

```
cos(alpha)
```

Let α be the input value. We define

$$\begin{aligned} f(\alpha) &= \cos(\alpha), \\ \frac{\partial f(\alpha)}{\partial x_i} &= -\sin(\alpha)\frac{\partial \alpha}{\partial x_i}, \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= -\cos(\alpha)\frac{\partial \alpha}{\partial x_i}\frac{\partial \alpha}{\partial x_j} - \sin(\alpha)\frac{\partial^2 \alpha}{\partial x_i \partial x_j}. \end{aligned}$$

Derive This expression calculates the derivative of the input with respect to one of the literals. If `alpha` is the input expression, the syntax for its derivative with respect to a literal `y` is

```
Derive(alpha, 'y')
```

Let α be the input value. Then,

$$f(\alpha) = \nu\left(\frac{\partial \alpha}{\partial y}\right).$$

The derivatives of this expression are not evaluated by Biogeme.

Integrate This expression performs numerical integration using a Gauss-Hermite algorithm. If takes as argument an expression `alpha` that includes a random variable `omega`. The syntax is

```
Integrate(alpha, 'omega')
```

Let α be the input value. We define

$$\begin{aligned} f(\alpha) &= \nu\left(\int_{-\infty}^{+\infty} \alpha(\omega) d\omega\right), \\ \frac{\partial f(\alpha)}{\partial x_i} &= \nu\left(\int_{-\infty}^{+\infty} \frac{\partial \alpha(\omega)}{\partial x_i} d\omega\right), \\ \frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= \nu\left(\int_{-\infty}^{+\infty} \frac{\partial^2 \alpha(\omega)}{\partial x_i \partial x_j} d\omega\right). \end{aligned}$$

NEED TO INCLUDE MORE INFO ABOUT THE IMPLEMENTATION

MonteCarlo This expression approximates an integral using Monte-Carlo integration. If takes as argument an expression `alpha` that includes draws. The syntax is

```
MonteCarlo(alpha)
```

Let α be the input value. We define

$$
\begin{aligned}
f(\alpha) &= \nu\left(\frac{1}{R}\sum_{r=1}^{R}\alpha(\xi_r)\right), \\
\frac{\partial f(\alpha)}{\partial x_i} &= \nu\left(\frac{1}{R}\sum_{r=1}^{R}\frac{\partial \alpha(\xi_r)}{\partial x_i}\right), \\
\frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= \nu\left(\frac{1}{R}\sum_{r=1}^{R}\frac{\partial^2 \alpha(\xi_r)}{\partial x_i \partial x_j}\right),
\end{aligned}
$$

where R is a parameter defining the number of draws to be used, and ξ_r are the values of the draws.

bioNormalCdf This expression provides an analytical approximation of the cumulative distribution function of a normal random variable. If `alpha` is the input, the syntax is

```
bioNormalCdf(alpha).
```

Let α be the input value. We define

$$
\begin{aligned}
f(\alpha) &=, \\
\frac{\partial f(\alpha)}{\partial x_i} &=, \\
\frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= .
\end{aligned}
$$

NEED TO INCLUDE MORE INFO ABOUT THE IMPLEMENTATION

Belongs to This expression verifies if the input value belongs to a set. If `alpha` is the input, the syntax is

```
BelongsTo(alpha, {1, 2, 3})
```

It returns the value 1 if the value belongs to the set, and 0 otherwise. The function is not differentiable.

Trajectory This expression is necessary when the data is organized as panel data. It means that several observations are available for the same individual. If `alpha` is the input, the syntax is

```
PanelLikelihoodTrajectory(alpha)
```

We denote by α_t the value of the input expression for observation t. Those values must be positive. If not, an exception is raised.

Then, we define

$$
\begin{aligned}
f(\alpha) &= \nu\left(\prod_{t=1}^{T}\alpha_t\right),\\
\frac{\partial f(\alpha)}{\partial x_i} &= \nu\left(f(\alpha)\sum_{t=1}^{T}\frac{1}{\alpha_t}\frac{\partial \alpha_t}{\partial x_i}\right),\\
\frac{\partial^2 f(\alpha)}{\partial x_i \partial x_j} &= \nu\left(\frac{1}{f(\alpha)}\frac{\partial f(\alpha)}{\partial x_i}\frac{\partial f(\alpha)}{\partial x_j} + f\sum_{t=1}^{T}\left[\frac{1}{\alpha_t}\frac{\partial^2 \alpha_t}{\partial x_i \partial x_j} - \frac{1}{\alpha_t^2}\frac{\partial \alpha_t}{\partial x_i}\frac{\partial \alpha_t}{\partial x_j}\right]\right).
\end{aligned}
$$

3.3 Comparison expressions

A comparison expression expects two expressions as argument. The output is either 0 or 1, where 0 means "False" and 1 means "True".

Equal This expression returns 1 if the two arguments have the same value, and 0 otherwise. The expression is not differentiable. If `alpha` and `beta` are the two arguments, the syntax is

```
alpha == beta
```

Not equal This expression returns 1 if the two arguments do not have the same value, and 0 otherwise. The expression is not differentiable. If `alpha` and `beta` are the two arguments, the syntax is

```
alpha != beta
```

Greater than This expression returns 1 if the value of the first argument is strictly greater than the value of the second one, and 0 otherwise. The expression is not differentiable. If `alpha` and `beta` are the two arguments, the syntax is

```
alpha > beta
```

Greater or equal than This expression returns 1 if the value of the first argument is greater or equal than the value of the second one, and 0 otherwise. The expression is not differentiable. If `alpha` and `beta` are the two arguments, the syntax is

```
alpha >= beta
```

bioExprLess.h bioExprLessOrEqual.h

3.4 Binary expressions

bioExprPlus.h bioExprMinus.h bioExprTimes.h bioExprDivide.h bioExprPower.h bioExprMin.h bioExprMax.h bioExprAnd.h bioExprOr.h

3.5 n-ary expressions

bioExprMultSum.h bioExprElem.h bioExprLinearUtility.h

3.6 Logit expressions

bioExprLogLogit.h bioExprLogLogitFullChoiceSet.h

References

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