### SSJ User's Guide

Package functionfit

Function fit utilities

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This package provides basic facilities for curve fitting and interpolation with polynomials as, for example, least square fit and spline interpolation.

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### PolInterp

Represents a polynomial that interpolates through a set of points. More specifically, let  $(x_0, y_0), \ldots, (x_n, y_n)$  be a set of points and p(x) the constructed polynomial of degree n. Then, for  $i = 0, \ldots, n$ ,  $p(x_i) = y_i$ .

```
package umontreal.iro.lecuyer.functionfit;
public class PolInterp extends Polynomial implements Serializable
Constructors
   public PolInterp (double[] x, double[] y)
      Constructs a new polynomial interpolating through the given points (x[0], y[0]), ...,
      (x[n], y[n]). This constructs a polynomial of degree n from n+1 points.
Methods
   public static double[] getCoefficients (double[] x, double[] y)
      Computes and returns the coefficients the polynomial interpolating through the given points
      (x[0], y[0]), \ldots, (x[n], y[n]). This polynomial has degree n and there are n+1 co-
      efficients.
   public double[] getX()
      Returns the x coordinates of the interpolated points.
   public double[] getY()
      Returns the y coordinates of the interpolated points.
   public static String toString (double[] x, double[] y)
      Makes a string representation of a set of points.
```

Calls toString (double[], double[]) with the associated points.

public String toString()

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# LeastSquares

This class implements different linear regression models, using the least squares method to estimate the regression coefficients. Given input data  $x_{ij}$  and response  $y_i$ , one want to find the coefficients  $\beta_i$  that minimize the residuals of the form (using matrix notation)

$$r = \min_{\beta} \|Y - X\beta\|_2,$$

where the  $L_2$  norm is used. Particular cases are

$$r = \min_{\beta} \sum_{i} \left( y_i - \beta_0 - \sum_{j=1}^{k} \beta_j x_{ij} \right)^2.$$

for k regressor variables  $x_i$ . The well-known case of the single variable x is

$$r = \min_{\alpha, \beta} \sum_{i} (y_i - \alpha - \beta x_i)^2.$$

Sometimes, one wants to use a basis of general functions  $\psi_j(t)$  with a minimization of the form

$$r = \min_{\beta} \sum_{i} \left( y_i - \sum_{j=1}^{k} \beta_j \psi_j(t_i) \right)^2.$$

For example, we could have  $\psi_j(t) = e^{-\lambda_j t}$  or some other functions. In that case, one has to choose the points  $t_i$  at which to compute the basis functions, and use a method below with  $x_{ij} = \psi_j(t_i)$ .

package umontreal.iro.lecuyer.functionfit;

public class LeastSquares

#### Methods

public static double[] calcCoefficients (double[] X, double[] Y)

Computes the regression coefficients using the least squares method. This is a simple linear regression with 2 regression coefficients,  $\alpha$  and  $\beta$ . The model is

$$y = \alpha + \beta x$$
.

Given the *n* data points  $(X_i, Y_i)$ ,  $i = 0, 1, \ldots, (n-1)$ , the method computes and returns the array  $[\alpha, \beta]$ .

public static double[] calcCoefficients (double[] X, double[] Y, int deg) Computes the regression coefficients using the least squares method. This is a linear regression with a polynomial of degree deg = k and k+1 regression coefficients  $\beta_j$ . The model is

$$y = \beta_0 + \sum_{j=1}^k \beta_j x^j.$$

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Given the *n* data points  $(X_i, Y_i)$ ,  $i = 0, 1, \ldots, (n-1)$ , the method computes and returns the array  $[\beta_0, \beta_1, \ldots, \beta_k]$ . Restriction: n > k.

### public static double[] calcCoefficientsO (double[][] X, double[] Y)

Computes the regression coefficients using the least squares method. This is a model for multiple linear regression. There are k+1 regression coefficients  $\beta_j$ , and k regressors variables  $x_j$ . The model is

$$y = \beta_0 + \sum_{j=1}^k \beta_j x_j.$$

There are n data points  $Y_i$ ,  $X_{ij}$ ,  $i=0,1,\ldots,(n-1)$ , and each  $X_i$  is a k-dimensional point. Given the response Y[i] and the regressor variables X[i][j],  $i=0,1,\ldots,(n-1)$ ,  $j=0,1,\ldots,(k-1)$ , the method computes and returns the array  $[\beta_0,\beta_1,\ldots,\beta_k]$ . Restriction: n>k+1.

### public static double[] calcCoefficients (double[][] X, double[] Y)

Computes the regression coefficients using the least squares method. This is a model for multiple linear regression. There are k regression coefficients  $\beta_j$ ,  $j=0,1,\ldots,(k-1)$  and k regressors variables  $x_j$ . The model is

$$y = \sum_{j=0}^{k-1} \beta_j x_j.$$

There are n data points  $Y_i$ ,  $X_{ij}$ ,  $i=0,1,\ldots,(n-1)$ , and each  $X_i$  is a k-dimensional point. Given the response Y[i] and the regressor variables X[i][j],  $i=0,1,\ldots,(n-1)$ ,  $j=0,1,\ldots,(k-1)$ , the method computes and returns the array  $[\beta_0,\beta_1,\ldots,\beta_{k-1}]$ . Restriction: n>k.

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# **BSpline**

Represents a B-spline with control points at  $(X_i, Y_i)$ . Let  $\mathbf{P_i} = (X_i, Y_i)$ , for  $i = 0, \dots, n-1$ , be a *control point* and let  $t_j$ , for  $j = 0, \dots, m-1$  be a *knot*. A B-spline [1] of degree p = m - n - 1 is a parametric curve defined as

$$\mathbf{P(t)} = \sum_{i=0}^{n-1} N_{i,p}(t) \mathbf{P_i}, \text{ for } t_p \le t \le t_{m-p-1}.$$

Here,

$$N_{i,p}(t) = \frac{t - t_i}{t_{i+p} - t_i} N_{i,p-1}(t) + \frac{t_{i+p+1} - t}{t_{i+p+1} - t_{i+1}} N_{i+1,p-1}(t)$$

$$N_{i,0}(t) = \begin{cases} 1 & \text{for } t_i \leq t \leq t_{i+1}, \\ 0 & \text{elsewhere.} \end{cases}$$

This class provides methods to evaluate  $\mathbf{P}(\mathbf{t}) = (X(t), Y(t))$  at any value of t, for a B-spline of any degree  $p \geq 1$ . Note that the evaluate method of this class can be slow, since it uses a root finder to determine the value of  $t^*$  for which  $X(t^*) = x$  before it computes  $Y(t^*)$ .

package umontreal.iro.lecuyer.functionfit;

public class BSpline implements MathFunction

#### Constructors

public BSpline (final double[] x, final double[] y, final int degree)

Constructs a new uniform B-spline of degree degree with control points at (x[i], y[i]). The knots of the resulting B-spline are set uniformly from x[0] to x[n-1].

public BSpline (final double[] x, final double[] y, final double[] knots)

Constructs a new uniform B-spline with control points at (x[i], y[i]), and knot vector given by the array knots.

#### Methods

public double[] getX()

Returns the  $X_i$  coordinates for this spline.

public double[] getY()

Returns the  $Y_i$  coordinates for this spline.

public double getMaxKnot()

Returns the knot maximal value.

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public double getMinKnot()

Returns the knot minimal value.

public double[] getKnots()

Returns an array containing the knot vector  $(t_0, t_{m-1})$ .

Returns a B-spline curve of degree degree interpolating the  $(x_i, y_i)$  points [1]. This method uses the uniformly spaced method for interpolating points with a B-spline curve, and a uniformed clamped knot vector, as described in http://www.cs.mtu.edu/~shene/COURSES/cs3621/NOTES/.

Returns a B-spline curve of degree degree smoothing  $(x_i, y_i)$ , for i = 0, ..., n points. The precision depends on the parameter h:  $1 \le \text{degree} \le h < n$ , which represents the number of control points used by the new B-spline curve, minimizing the quadratic error

$$L = \sum_{i=0}^{n} \left( \frac{Y_i - S_i(X_i)}{W_i} \right)^2.$$

This method uses the uniformly spaced method for interpolating points with a B-spline curve and a uniformed clamped knot vector, as described in http://www.cs.mtu.edu/~shene/COURSES/cs3621/NOTES/.

#### public BSpline derivativeBSpline()

Returns the derivative B-spline object of the current variable. Using this function and the returned object, instead of the derivative method, is strongly recommended if one wants to compute many derivative values.

#### public BSpline derivativeBSpline (int i)

Returns the *i*th derivative B-spline object of the current variable; *i* must be less than the degree of the original B-spline. Using this function and the returned object, instead of the derivative method, is strongly recommended if one wants to compute many derivative values.

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# SmoothingCubicSpline

Represents a cubic spline with nodes at  $(x_i, y_i)$  computed with the smoothing cubic spline algorithm of Schoenberg [1, 2]. A smoothing cubic spline is made of n+1 cubic polynomials. The *i*th polynomial of such a spline, for  $i=1,\ldots,n-1$ , is defined as  $S_i(x)$  while the complete spline is defined as

$$S(x) = S_i(x), \qquad \text{for } x \in [x_{i-1}, x_i].$$

For  $x < x_0$  and  $x > x_{n-1}$ , the spline is not precisely defined, but this class performs extrapolation by using  $S_0$  and  $S_n$  linear polynomials. The algorithm which calculates the smoothing spline is a generalization of the algorithm for an interpolating spline.  $S_i$  is linked to  $S_{i+1}$  at  $x_{i+1}$  and keeps continuity properties for first and second derivatives at this point, therefore  $S_i(x_{i+1}) = S_{i+1}(x_{i+1})$ ,  $S_i'(x_{i+1}) = S_{i+1}'(x_{i+1})$  and  $S_i''(x_{i+1}) = S_{i+1}''(x_{i+1})$ .

The spline is computed with a smoothing parameter  $\rho \in [0, 1]$  which represents its accuracy with respect to the initial  $(x_i, y_i)$  nodes. The smoothing spline minimizes

$$L = \rho \sum_{i=0}^{n-1} w_i (y_i - S_i(x_i))^2 + (1 - \rho) \int_{x_0}^{x_{n-1}} (S''(x))^2 dx$$

In fact, by setting  $\rho = 1$ , we obtain the interpolating spline; and we obtain a linear function by setting  $\rho = 0$ . The weights  $w_i > 0$ , which default to 1, can be used to change the contribution of each point in the error term. A large value  $w_i$  will give a large weight to the *i*th point, so the spline will pass closer to it. Here is a small example that uses smoothing splines:

```
package umontreal.iro.lecuyer.functionfit;
   import umontreal.iro.lecuyer.functions.*;
   import umontreal.iro.lecuyer.functions.Polynomial;
public class SmoothingCubicSpline implements MathFunction,
             MathFunctionWithFirstDerivative, MathFunctionWithDerivative,
             MathFunctionWithIntegral
```

#### Constructors

```
public SmoothingCubicSpline (double[] x, double[] y, double[] w,
                             double rho)
```

Constructs a spline with nodes at  $(x_i, y_i)$ , with weights  $w_i$  and smoothing factor  $\rho = \text{rho}$ . The  $x_i$  must be sorted in increasing order.

```
public SmoothingCubicSpline (double[] x, double[] y, double rho)
```

Constructs a spline with nodes at  $(x_i, y_i)$ , with weights = 1 and smoothing factor  $\rho = \text{rho}$ . The  $x_i$  must be sorted in increasing order.

#### Methods

```
public double evaluate (double z)
```

Evaluates and returns the value of the spline at z.

```
public double integral (double a, double b)
```

Evaluates and returns the value of the integral of the spline from a to b.

```
public double derivative (double z)
```

Evaluates and returns the value of the first derivative of the spline at z.

```
public double derivative (double z, int n)
```

Evaluates and returns the value of the n-th derivative of the spline at z.

```
public double[] getX()
```

Returns the  $x_i$  coordinates for this spline.

```
public double[] getY()
```

Returns the  $y_i$  coordinates for this spline.

```
public double[] getWeights()
```

Returns the weights of the points.

```
public double getRho()
```

Returns the smoothing factor used to construct the spline.

```
public Polynomial[] getSplinePolynomials()
```

Returns a table containing all fitting polynomials.

### public int getFitPolynomialIndex (double x)

Returns the index of P, the Polynomial instance used to evaluate x, in an ArrayList table instance returned by getSplinePolynomials(). This index k gives also the interval in table **X** which contains the value x (i.e. such that  $x_k < x \le x_{k+1}$ ).

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### References

[1] C. de Boor. A Practical Guide to Splines. Number 27 in Applied Mathematical Sciences Series. Springer-Verlag, New York, 1978.

[2] D. S. G. Pollock. Smoothing with cubic splines. Technical report, University of London, Queen Mary and Westfield College, London, 1993.