OpenTURNS cheat sheet



This OpenTURNS v1.17 cheat sheet provides a quick overview of all the programming interface. For full documentation, please read the doc. A beginner may be interested in the Quick start guides.

This cheat sheet follows the steps of the ABC method.

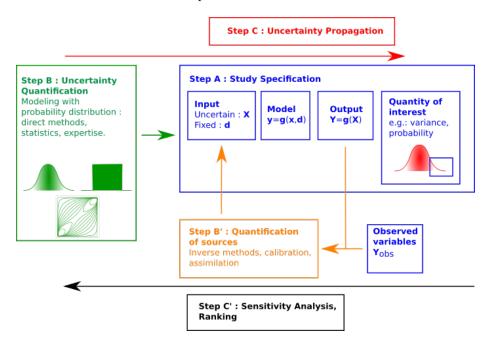


Figure 1: The ABC method

Step A: define the study

Purpose	Class / Method
Import OpenTURNS	import openturs as ot
Vector	ot.Point(dimension)
Sample	ot.Sample(size, dimension)
Symbolic function	ot.SymbolicFunction(["x0",
	"x1"], ["1 + x0 + x1"])
Python function	ot.PythonFunction(number_of_inputs,
	<pre>number_of_outputs, g)</pre>
Manage history and cache	ot.MemoizeFunction(myfunction)

Purpose	Class / Method
Normal	ot.Normal(mu, sigma)
Uniform	ot.Uniform(a, b)
Multivariate distribution with	<pre>ot.ComposedDistribution((X0,</pre>
independent copula	X1, X2))
Input random vector	ot.RandomVector(inputDistribution)
Output random vector	ot.CompositeRandomVector(g,
	inputRandomVector)
Generate observations	randomVector.getSample(sample_size)
Set the seed	ot.RandomGenerator.SetSeed(1976)
Get sample size	<pre>sample.getSize()</pre>
Get sample dimension	<pre>sample.getDimension()</pre>
Compute sample mean	outputSample.computeMean()
Compute sample standard deviation	outputSample.computeStandardDeviation()

Step B : quantification of the sources of uncertainties

Purpose	Class / Method
Fit a Normal	ot.NormalFactory().build(sample)
Fit a Beta	ot.BetaFactory().build(sample)
Fit an histogram	ot.HistogramFactory().build(sample
Fit a kernel density estimator	<pre>ot.KernelSmoothing().build(sample)</pre>
Draw QQ-plot	ot.VisualTest.DrawQQplot(sample,
	distribution)
Kolmogorov-Smirnov test (known	<pre>ot.FittingTest.Kolmogorov(sample,</pre>
parameters)	distribution)
Kolmogorov-Smirnov test (unknown	ot.FittingTest.Lilliefors(sample,
parameters)	factory)
BIC criteria	ot.FittingTest.BIC(sample,
	distribution)

Step C : push forward the uncertainties

Purpose	Class / Method
Taylor expansion	ot.TaylorExpansionMoments(output_random_vector)
Estimate mean	ot.ExpectationSimulationAlgorithm(output_random_vector)
Estimate $P(Y > s)$	sample.computeEmpiricalCDF(s,
, ,	True)
Create the event $(Y > s)$	<pre>ot.ThresholdEvent(output_random_vector,</pre>
,	ot.Greater(), s)
Create a Monte-Carlo experiment	ot.MonteCarloExperiment()

Purpose	Class / Method
Estimate a probability	<pre>ot.ProbabilitySimulationAlgorithm(myEvent, experiment)</pre>

Step C': sensitivity analysis

Purpose	Class / Method
Perform linear regression	<pre>ot.LinearLeastSquares(input_sample, output_sample)</pre>
Standardized regression coefficients	<pre>ot.CorrelationAnalysis_SignedSRC(input_sample, output_sample)</pre>
Draw indices	<pre>ot.SobolIndicesAlgorithm.DrawCorrelationCoefficients(src_ input_names, "SRC coefficients")</pre>
Estimate Sobol' indices given budget	<pre>ot.SobolIndicesExperiment(distribution, sample_size)</pre>
Estimate Sobol' indices	ot.SaltelliSensitivityAlgorithm()

Step B': calibration

Purpose	Class / Method
Create the parametric model	ot.ParametricFunction(g,
	<pre>calibrated_indices,</pre>
	theta_prior)
Linear least squares	ot.LinearLeastSquaresCalibration(parametric_g,
	<pre>input_sample, output_sample,</pre>
	theta_prior, "SVD")
Non linear least squares	$\verb ot.NonLinearLeastSquaresCalibration(parametric_g,$
	<pre>input_sample, output_sample,</pre>
	theta_prior)
Linear gaussian	ot.GaussianLinearCalibration(parametric_g,
	<pre>input_sample, output_sample,</pre>
	theta_prior, theta_sigma,
	<pre>output_covariance)</pre>
Non linear gaussian	$\verb ot.GaussianNonLinearCalibration(parametric_g, \\$
	<pre>input_sample, output_sample,</pre>
	theta_prior, theta_sigma,
	<pre>output_covariance)</pre>

Purpose	Class / Method
Bayesian calibration	ot.RandomWalkMetropolisHastings(prior, conditional, model, input_sample, output_sample, initialState, proposal)

Metamodel

Purpose	Class / Method
Squared exponential	ot.SquaredExponential([1.0] *
	dimension, [1.0])
Matern 5/2 covariance	ot.MaternModel([1.0] *
	dimension, 2.5)
Kriging	ot.KrigingAlgorithm(input_sample,
	output_sample, covarianceModel,
	basis)
Sample from kriging	ot.KrigingRandomVector(result,
	input_sample)
Conditioned gaussian process	ot.ConditionedGaussianProcess(kriging_result,
	mesh)
Multivariate basis	ot.OrthogonalProductPolynomialFactory(distribution_colle
Polynomial chaos (given data)	ot.FunctionalChaosAlgorithm(input_sample,
,	output_sample)
Polynomial chaos (given distribution)	ot.FunctionalChaosAlgorithm(input_sample,
,	output_sample, distribution,
	adaptive_strategy,
	projection_strategy)
Sobol' indices from chaos	ot.FunctionalChaosSobolIndices(functional_chaos_result)
Sample from chaos	ot.FunctionalChaosRandomVector(functional_chaos_result)
Validation	ot.MetaModelValidation(input_test,
	output_test, metamodel)

Design of experiments

Purpose	Class / Method
Monte-Carlo	<pre>ot.MonteCarloExperiment(distribution, sample_size)</pre>
Latin Hypercube Sampling	<pre>ot.LHSExperiment(distribution, sample_size)</pre>
Optimized LHS	ot.SimulatedAnnealingLHS(lhs_experiment, criteria, temperature_profile)
Sobol' sequence	ot.SobolSequence(dimension)

Purpose	Class / Method
Low discrepancy sequence	ot.LowDiscrepancyExperiment(ld_sequence, distribution, samplesize,
Import viewer Plot DOE	<pre>False) import openturs.viewer as otv otv.PlotDesign(sample, bounds)</pre>

More resources

Resource	Link
Forum	https://openturns.discourse.group
Chat	https://gitter.im/openturns/community
Modules	https://github.com/openturns/openturns/wiki/Modules
Install	http://openturns.github.io/openturns/master/install.html
Bugs	https://github.com/openturns/openturns/issues
Events	https://github.com/openturns/openturns/wiki/OpenTURNS-events
Bibliography	https://github.com/openturns/openturns/wiki/Bibliography
Bib resources	Bibtex file
Presentations	https://github.com/openturns/presentation