

Decision Variables' Forecast Statistic

Asset 1
Number of Datapoints 20,000
Mean 0.3065
Median 0.3066
Standard Deviation 0.0022
Variance 0.0000
Average Deviation 0.0017
Maximum 0.3105
Minimum 0.3030
Range 0.0074
Skewness 0.0824
Kurtosis -0.7742
25% Percentile 0.3042
75% Percentile 0.3076
Error Precision at 95% 0.0031



Asset 2
Number of Datapoints 20,000
Mean 0.1931
Median 0.1933
Standard Deviation 0.0024
Variance 0.0000
Average Deviation 0.0020
Maximum 0.1965
Minimum 0.1887
Range 0.0077
Skewness -0.2548
Kurtosis -1.1475
25% Percentile 0.1910
75% Percentile 0.1950
Error Precision at 95% 0.0043



Asset 3
Number of Datapoints 20,000
Mean 0.1
Median 0.1
Standard Deviation 0.0019
Variance 0.0000
Average Deviation 0.0019
Maximum 0.1019
Minimum 0.0981
Range 0.038
Skewness 0.0019
Kurtosis 0.0019
25% Percentile 0.0981
75% Percentile 0.1019
Error Precision at 95% 0.0019

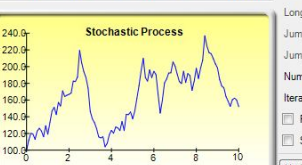
Stochastic Processes are sequences of events or paths generated by probabilistic laws where random events can occur over time but are governed by specific statistical and probabilistic rules. They are useful for forecasting random events (e.g., stock prices, interest rates, price of electricity).

Genetic Algorithm

Genetic
Objective Cell:
Variables: Add
ColumnCell

Constraints: Add
ColumnCell

Methods
☒ Brownian Motion (Random Walk) with Drift
☐ Exponential Brownian Motion (Random Walk) with Drift
☐ Mean-Reversion Process with Drift
☐ Jump-Diffusion Process with Drift
☐ Jump-Diffusion Process with Drift and Mean-Reversion



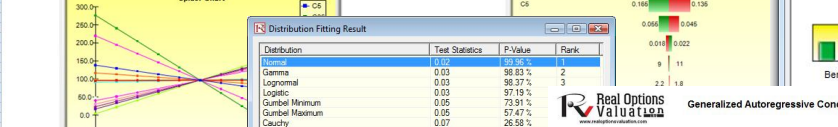
Statistical Summary

One of the powerful simulation tools is the tornado chart. It captures the static impacts of each variable on the outcome of the model. That is, the tool examines many periods each precedent variable in the model a user-specified preset amount, captures the fluctuation on the model's forecast or final result, and lists the resulting perturbations ranked from the most significant to the least. Precedents are all the input and intermediate variables that affect the outcome of the model. For instance, if the model consists of $A = B + C$, where $C = D + E$, then B, D, and E are the precedents for A (C is not a precedent as it is only an intermediate calculated value). The range and number of values perturbed is user-specified and can be set to test extreme values rather than smaller perturbations for A (C is not the expected values). In certain circumstances, extreme values may have a larger, smaller, or unbalanced impact (e.g., nonlinearities may occur where increasing or decreasing economies of scale and scope creep occurs for larger or smaller values of a variable) and only a wider range will capture this nonlinear impact.

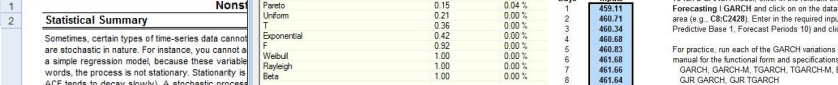
A tornado chart lists all the inputs that drive the model, starting from the input variable that has the most effect on the results. The chart is obtained by perturbing each precedent input at some consistent range (e.g., $\pm 10\%$ from the base case) one at a time, and comparing their results to the base case. A spider chart looks like a spider with a central body and its many legs protruding. The positively sloped line indicates a positive relationship, while a negatively sloped line indicates a negative relationship. Further, spider charts can be used to visualize linear and nonlinear relationships. The tornado and spider charts help identify the critical success factors of an output cell in order to identify the inputs to simulate. The identified critical variables that are uncertain are the ones that should be simulated. Do not waste time simulating variables that are neither uncertain nor have little impact on the results.

Precedent Cell	Base Value: 96.628183551219	Input Charges	Output	Output	Effective	Input	Input	CSSE	Value
			Downside	Upside	Range	Downside	Upside		
C36 C36	270.6282	83.37384	160.00	380.00	220.00	160.00	380.00	1.0000	1.0000
C36 C9	219.7269	-26.4746	246.20	200.00	36.00%	44.00%	40.00%	0.00%	0.00%
C12 C12	34.22542	189.8268	186.40	59.00	59.00	111.00	51.00	0.00%	0.00%
C13 C13	10.70603	176.5457	169.84	81.13	81.13	13.48	12.25	0.00%	0.00%
C15 C15	23.1775	170.0748	146.90	45.00	55.00	55.00	38.00	0.00%	0.00%
C16 C16	30.533	162.7193	132.18	31.50	38.50	38.50	35.00	0.00%	0.00%
C14 C14	40.14659	153.1057	112.95	81.34	81.34	115.67	81.34	0.00%	0.00%
C17 C17	48.04737	145.205	97.16	18.00	22.00	20.00	20.00	0.00%	0.00%
C5 C5	138.2391	57.02984	81.21	13.50%	16.50%	15.00%	0.00%	0.00%	0.00%
C8 C8	116.8038	75.64095	40.16	4.50%	5.50%	5.00%	0.00%	0.00%	0.00%
C7 C7	90.58835	102.6854	12.10	1.80%	2.20%	2.00%	0.00%	0.00%	0.00%
C24 C24	95.08417	98.16816	3.08	59.00	111.00	51.00	0.00%	0.00%	0.00%
C27 C27	97.08876	96.16357	0.93	\$1.00	\$2.00	\$2.00	0.00%	0.00%	0.00%
C25 C25	96.16357	97.08876	0.93	\$2.70	\$3.30	\$3.00	0.00%	0.00%	0.00%

Tornado Chart



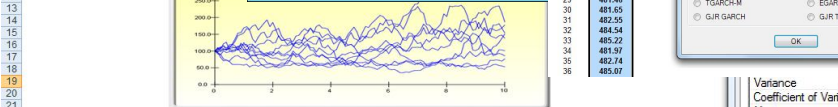
Spider Chart



Distribution Fitting Result

Distribution	Test Statistics	P-Value	Rank
Gamma	0.03	98.93 %	2
Lognormal	0.03	98.37 %	2
Logistic	0.03	97.10 %	2
Gumbel Minimum	0.05	73.91 %	3
Gumbel Maximum	0.05	57.47 %	3
Levy	0.07	26.58 %	4
Triangular	0.08	15.30 %	4
Chi-Square	0.10	3.11 %	5
Pareto	0.15	0.04 %	6
Uniform	0.21	0.00 %	7
Exponential	0.42	0.00 %	8
Weibull	0.50	0.00 %	9
Rayleigh	1.00	0.00 %	10
Beta	1.00	0.00 %	10

Theoretical vs. Empirical Distribution



Statistical Summary

The following are the estimated parameters for a stochastic process given the data provided. It is up to you to determine if the probability of fit (similar to a goodness-of-fit computation) is sufficient to warrant the use of a stochastic process forecast, and if so, whether it is a random walk, mean-reversion, or a jump-diffusion model, or combinations thereof. In choosing the right stochastic process model, you will have to rely on past experience and a prior economic and financial expectations of what the underlying data sets is best represented by. These parameters can be entered into a stochastic process forecast (Simulation Forecasting) Stochastic Processes).

Periodic	Drift Rate	-1.48%	Reversion Rate	283.89%	Jump Rate	20.41%
	Volatility	88.84%	Long-Term Value	327.72	Jump Size	237.89

Probability of stochastic model fit: 46.48%
A high fit means a stochastic model is better than conventional models.
Runs 20
Standard Normal -1.7321
Positive 29
P-Value (1-tail) 0.0416
Negative 25
P-Value (2-tail) 0.0833
Expected Run 26
A low p-value (below 0.10, 0.05, 0.01) means that the sequence is not random and hence suffer

Descriptive Statistics

require one moment, while others require two moments, and so forth). Descriptive statistics quantitatively sution (i.e., mean, median, and mode) and is interpreted as the expected value, expected returns, or the

up all of the data points and dividing them them to all be positive. The Geometric Mean growth rate given compound interest averages are prone to significant bias when

Optimization Summary

Optimization is used to allocate resources where the results provide the max returns or the min costs/risks. Uses including managing inventories, financial portfolio allocation, product mix, project selection, etc.

Objective Method Constraints Statistics Decision Variables

Static Optimization

Run on static model without simulations. Usually run to determine the initial optimal portfolio before more advanced optimizations are applied.

Dynamic Optimization

A simulation is first run, the results of the simulation are applied in the model, and then an optimization is applied to the simulated values.

Number of Simulation Trials 500

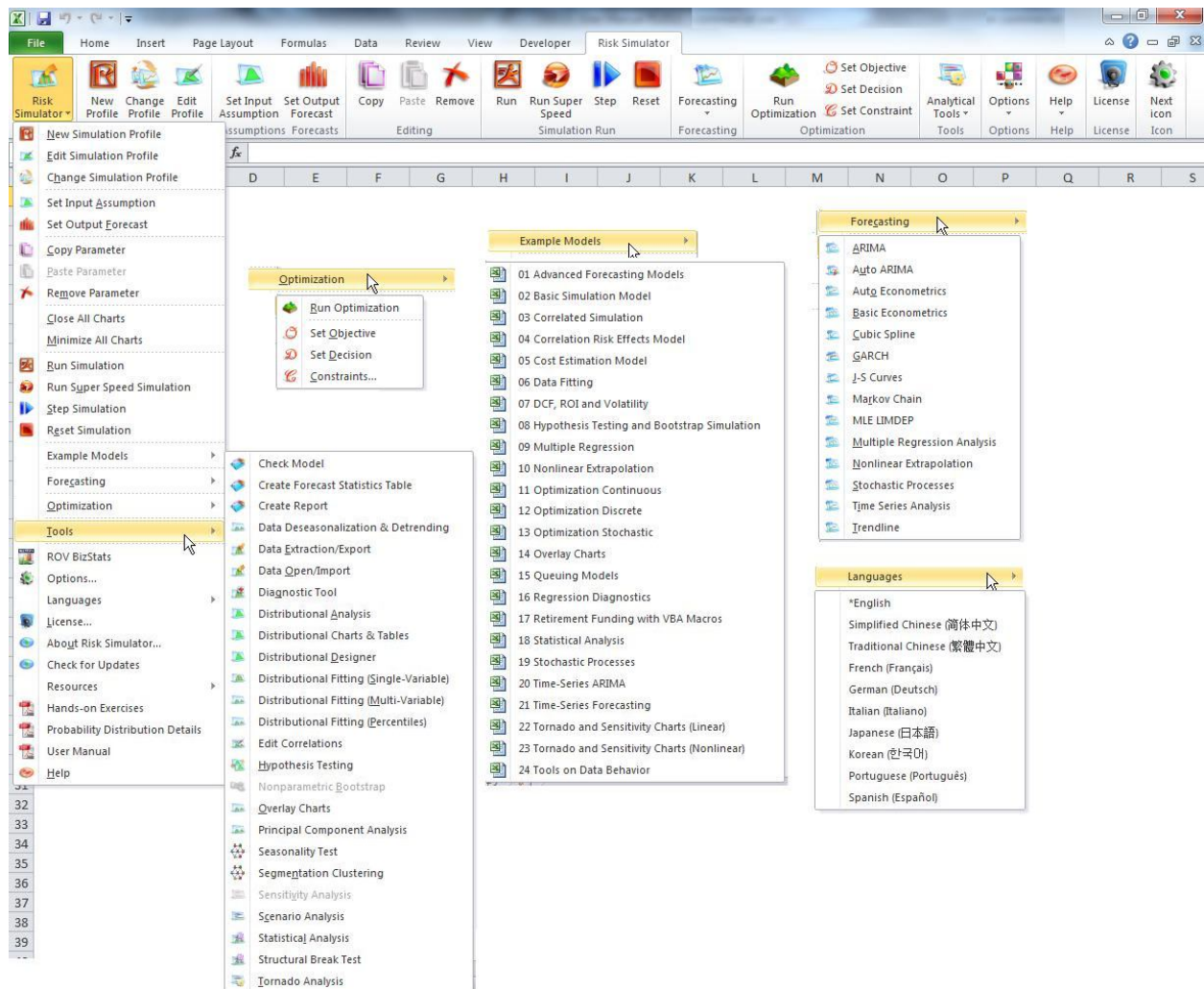
Stochastic Optimization

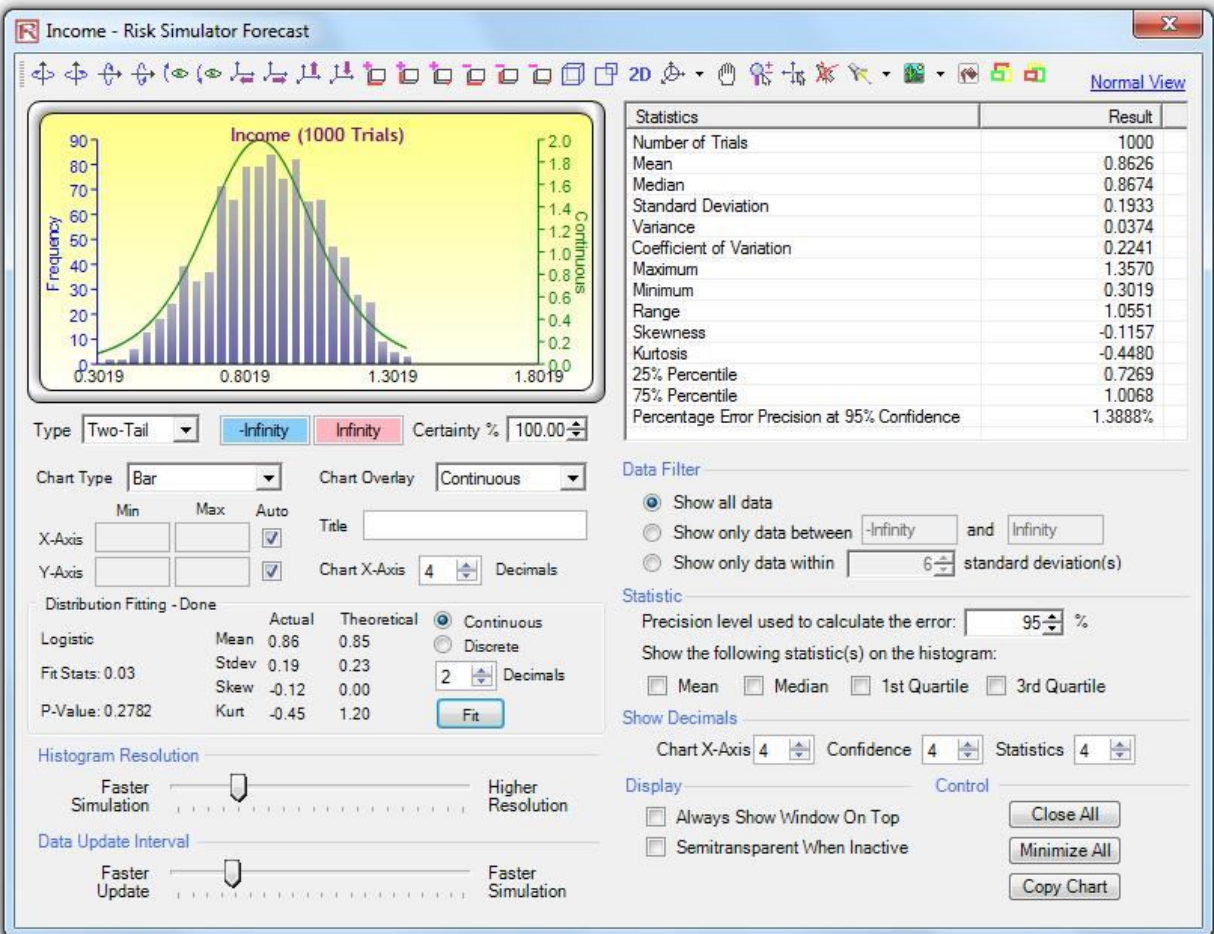
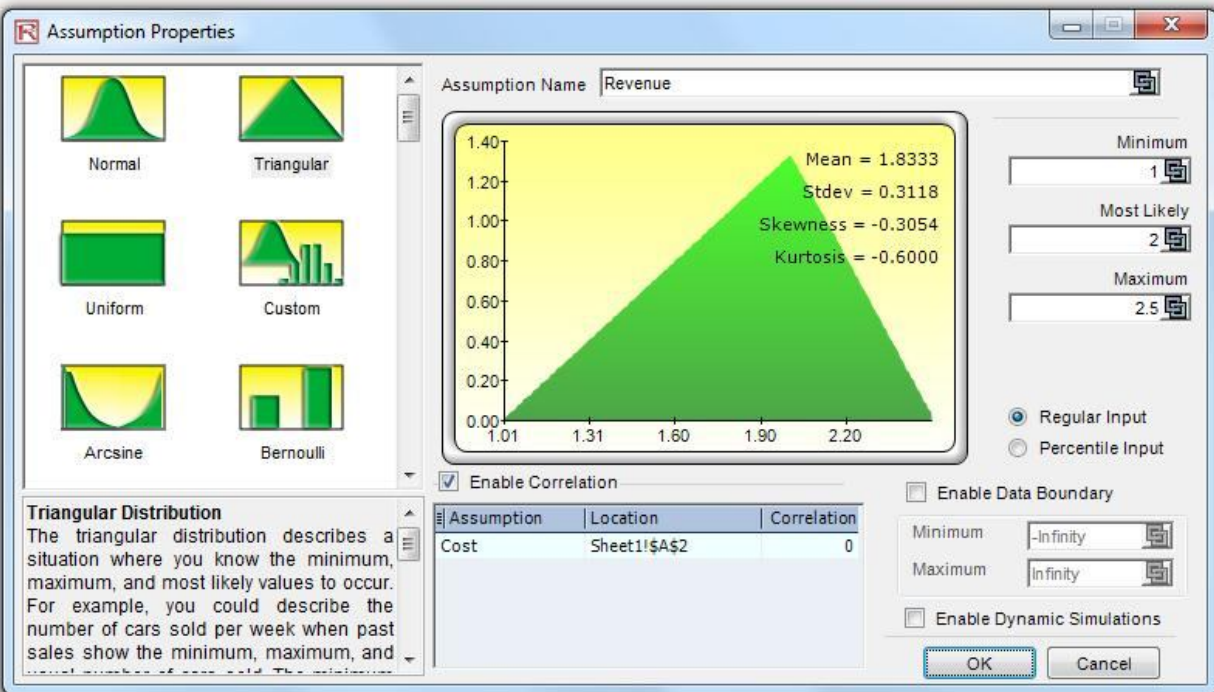
Similar to dynamic optimization but the process is repeated several times. The final decision variables will each have its own forecast chart indicating its optimal range.

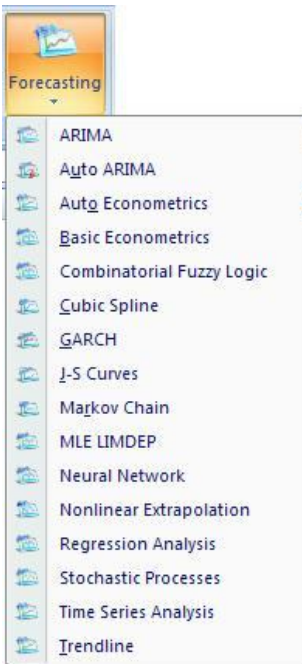
Number of Simulation Trials 5000

Number of Optimization Runs 100

Advanced OK Cancel







Stochastic Process Forecasting

Stochastic Processes are sequences of events or paths generated by probabilistic laws where random events can occur over time but are governed by specific statistical and probabilistic rules. They are useful for forecasting random events (e.g., stock prices, interest rates, price of electricity).

Methods

- ☒ Brownian Motion (Random Walk) with Drift
- ☐ Exponential Brownian Motion (Random Walk) with Drift
- ☐ Mean-Reversion Process with Drift
- ☐ Jump-Diffusion Process with Drift
- ☐ Jump-Diffusion Process with Drift and Mean-Reversion

Stochastic Process

Parameters:

Starting Value	100
Growth or Drift Rate (%)	5
Annualized Volatility (%)	25
Forecast Horizon (Years)	10
Reversion Rate (%)	5
Long-Term Value	120
Jump Rate (%)	10
Jump Size	2
Number of Steps	100
Iterations	10
☐ Random Seed	1
☐ Show All Iterations	

Update Chart OK Cancel

Binary Logistic Maximum Likelihood Forecast: Logit, Probit, Tobit

LOGIT & PROBIT SAMPLE DATA

Defaulted	Age	Education Level	Years with Current Employer	Years at Current Address	Household Income (Thousands \$)	Debt to Income Ratio (%)	Credit Card Debt (Thousands \$)	Other Debt (Thousands \$)
1	41	3	17	12	176	9.3	11.36	5.01
0	27	1	10	6	31	17.3	1.36	4
0	40	1	15	14	55	5.5	0.86	2.17
0	41	1	15	14	120	2.9	2.66	0.82
1	24	2	2	0	28	17.3	1.79	3.06
0	41	2	5	5	25	10.2	0.39	2.16
0	39	1	20	9	67	30.6	3.83	16.67
0	43	1	12	11	22	2.2	0.12	1.2
1	24	1	11	11	22	2.2	0.12	1.2
0	36	1	11	11	22	2.2	0.12	1.2
0	27	1	11	11	22	2.2	0.12	1.2
0	25	1	11	11	22	2.2	0.12	1.2
0	52	1	11	11	22	2.2	0.12	1.2
0	37	1	11	11	22	2.2	0.12	1.2
0	48	1	11	11	22	2.2	0.12	1.2
1	36	2	11	11	22	2.2	0.12	1.2
1	36	2	11	11	22	2.2	0.12	1.2
0	43	1	11	11	22	2.2	0.12	1.2
0	39	1	11	11	22	2.2	0.12	1.2
0	41	3	11	11	22	2.2	0.12	1.2
0	39	1	11	11	22	2.2	0.12	1.2
0	47	1	11	11	22	2.2	0.12	1.2
0	28	1	11	11	22	2.2	0.12	1.2
0	29	1	11	11	22	2.2	0.12	1.2
1	21	2	11	11	22	2.2	0.12	1.2
0	25	4	11	11	22	2.2	0.12	1.2
0	45	2	11	11	22	2.2	0.12	1.2
0	43	1	11	11	22	2.2	0.12	1.2
0	33	2	11	11	22	2.2	0.12	1.2
0	26	3	11	11	22	2.2	0.12	1.2
0	45	1	11	11	22	2.2	0.12	1.2
0	30	1	11	11	22	2.2	0.12	1.2

MLE LIMDEP

Runs the Logit, Probit and Tobit models for limited dependent variables (LIMDEP) where the dependent variable's (Y) data points are binary or limited to discrete values, and where the forecast of dependent values are probabilities of occurrences. In such situations, a regular regression analysis will yield incorrect and biased results, including the violation of normality requirements and forecast results of negative probabilities or values exceeding 100%. Only these LIMDEP models are appropriate for use when dependent variables are limited.

Dependent Variable
Defaulted

Defaulted	Age	Education Level
1	41	3
0	27	1
0	40	1
0	41	1
1	24	2
0	41	2
0	39	1
0	43	1
1	24	1

☒ Logit
☐ Probit
☐ Tobit

OK

Cancel

Basic Econometrics Data Set

Y	X1	X2	X3	X4	X5
521	18308	185	4.041	79.6	7.2
367	1148	600	0.55	1	8.5
443	18068	372	3.665	32.3	5.7
365	7729	142	2.351	45.1	7.3
614	100484	432	29.76	190.8	7.5
385	16728	290	3.294	31.8	5
286	14630	346	3.287	678.4	6.7
397	4008	328	0.666	340.8	6.2
764	38927	354	12.938	239.6	7.3
427	22322	266	6.478	111.9	5
153	3711	320	1.108	172.5	2.8
231	3136	197	1.007	12.2	6.1
524	50508	266	11.431	205.6	7.1
328	28886	173	5.544	154.6	5.9
240	16996	190	2.777	49.7	4.6
286	13035	239	2.478	30.3	4.4
285	12973	190	3.685	92.8	7.4
569	16309	241	4.22	96.9	7.1
96	5227	189	1.228	39.8	7.5
498	19235	358	4.781	489.2	5.9
481	44487	315	6.016	767.6	9
468	44213	303	9.295	163.6	9.2
177	23619	228	4.375	55	5.1
198	9106	134	2.573	54.9	8.6
458	24917	189	5.117	74.3	6.6
108	3872	196	0.799	5.5	6.9
246	8945	183	1.578	20.5	2.7
291	2373	417	1.202	10.9	5.5
68	7128	233	1.109	123.7	7.2
311	23624	349	7.73	1042	6.6
606	5242	284	1.515	12.5	6.9
512	92629	499	17.99	381	7.2
426	28795	231	6.629	136.1	5.8
47	4487	143	0.639	9.3	4.1
265	48799	249	10.847	264.9	6.4
370	14067	195	3.146	45.8	6.7
312	12693	288	2.842	29.6	6
222	62184	229	11.882	265.1	6.9
280	9153	287	1.003	960.3	8.5

Basic Econometrics

This tool is used to run basic econometric models by first transforming the input variables before running the multivariate regression analysis. You can enter in multiple econometric model specifications to test. Each model is on a new line and within each line, the first variable is the dependent variable followed by at least one or more independent variables separated by semi-colons. In the following example, LN(VAR1) and VAR3 are dependent variables in two models and the remaining items are independent variables in the two econometric models:

LN(VAR1); LN(VAR2); VAR3+VAR4; TIME
VAR3; LAG(VAR2,3); DIFF(VAR1); RESIDUAL(VAR3;VAR4)

VAR1	VAR2	VAR3	VAR4	VAR5	VAR6
521	18308	185	4.041	79.6	7.2
367	1148	600	0.55	1	8.5
443	18068	372	3.665	32.3	5.7
365	7729	142	2.351	45.1	7.3
614	100484	432	29.76	190.8	7.5
385	16728	290	3.294	31.8	5
286	14630	346	3.287	678.4	6.7

Single Model

Dependent Variable

LN(VAR1)

e.g., LN(VAR1)

Functions: +, -, *, /, LN, LOG, LAG

Independent Variables:

LN(VAR2); VAR3*VAR4; LAG(VAR5,1); DIFF(VAR6); TIME

e.g., LOG(VAR2+VAR3); VAR3*VAR4; LAG(VAR2); VAR5; RESIDUAL(VAR1;VAR3); TIME; FORECAST(VAR3;VAR4); DIFF(VAR5); RATE(VAR6)

Show Result

Multiple Models

INTEGER1: Min Max

INTEGER2: Min Max

INTEGER3: Min Max

Econometrics Results

R-Squared (Coefficient of Determination): 0.5231

Adjusted R-Squared: 0.4663

Multiple R (Multiple Correlation Coefficient): 0.7233

Standard Error of the Estimates (SE): 0.4666

ANOVA F Statistic: 9.2137

ANOVA p-Value: 0.0000

	Intercept	LN(VAR2)	VAR3*VAR4	LAG(VAR5,1)	DIFF(VAR6)	TIME
Coefficients	3.1049	0.2726	0.0000	0.0011	0.0219	-0.0125
Standard Error	0.8947	0.0974	0.0000	0.0003	0.0322	0.0049
t-Statistic	3.4703	2.8001	0.7885	3.8576	0.6796	-2.5234
p-Value	0.0012	0.0077	0.4348	0.0004	0.5005	0.0155

Dependent Variable:

LN(VAR1)

Copy

Close

Generalized Autoregressive Conditional Heteroskedasticity (GARCH)

Historical Data

Days	Inputs
1	459.11
2	460.71
3	460.34
4	460.68
5	460.83
6	461.68
7	461.66
8	461.64
9	465.97
10	469.38
11	470.05
12	469.72
13	466.95
14	464.78
15	465.81
16	465.86
17	467.44
18	468.32
19	470.39
20	468.51
21	470.42
22	470.4
23	472.78
24	478.64
25	481.14
26	480.81
27	481.19
28	480.19
29	481.46
30	481.65
31	482.55
32	484.54
33	485.22
34	481.97
35	482.74
36	485.07

To run a GARCH model, enter in the relevant time-series data, then click on **Risk Simulator | Forecasting | GARCH** and click on on the data location *link* icon, select the historical data area (e.g., C8:C2428). Enter in the required inputs (e.g., P 1, Q 1, Daily Trading Periodicity 252, Predictive Base 1, Forecast Periods 10) and click **OK**. Review the generated forecast report.

For practice, run each of the GARCH variations and compare the results. Refer to the user manual for the functional form and specifications for each model variation:

GARCH, GARCH-M, TGARCH, TGARCH-M, EGARCH, EGARCH-T
GJR GARCH, GJR TGARCH

GARCH
✕

GARCH or generalized autoregressive conditional heteroskedasticity models are used in forecasting the volatility of financial instruments, using the prices themselves. The GARCH (P,Q) model allows for different positive P and Q integer lag parameters for the mean (news) and variance equations. Note that only positive data values can be used in a GARCH volatility forecast. Periodicity is the number of periods per year (e.g., 12 for monthly data, 252 for daily trading data, 365 for daily data) to annualize the volatility or keep as 1 for periodic volatility. Base is the predictive base periods (this means how many periods back you would like to use as a forecast base to predict future volatility, e.g., enter in 12 if using the past 12 periods). Variance Targeting means if you wish the volatility forecast to revert to an imputed long-run mean over time. Make sure to arrange your raw price data in chronological order (past to present in a single column with multiple rows).

Data Location:

Generate a GARCH (P,Q) model for:

P: Q: Periodicity: Base: Forecast Periods:

☐ Apply Variance Targeting

☒ GARCH

☐ GARCH-M

☐ TGARCH

☐ TGARCH-M

☐ EGARCH

☐ EGARCH-T

☐ GJR GARCH

☐ GJR TGARCH

☐ Run All Models

OK

Cancel

Efficient Frontier

Problem Parameters:
Number of variables 12
Number of functions 3
Objective function will be Maximized

STEP1, D17 <= 5000, J17 <= 4

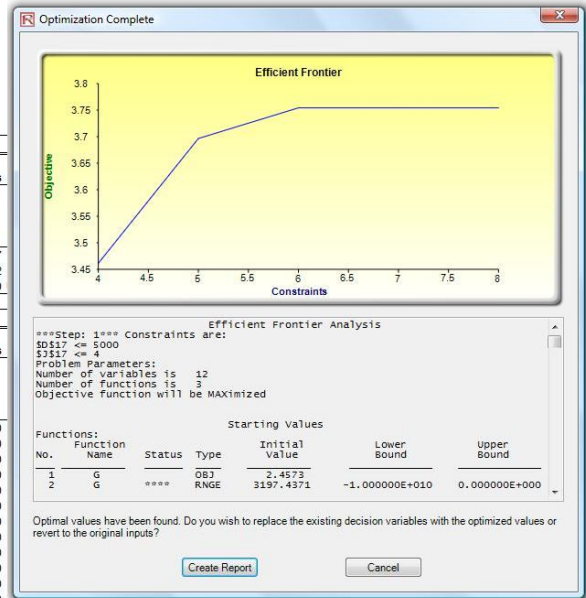
Functions

Starting Values							Final Results			
No.	Function Name	Status	Type	Initial Value	Lower Bound	Upper Bound	No.	Function Name	Initial Value	Final Value
1	G		OBJ	2.45726			1	G	2.45726	3.46137
2	G	****	RNGE	3197.43710	-1E+10	0	2	G	3197.43710	-1472.56292
3	G	****	RNGE	8.00000	-1E+10	0	3	G	8.00000	0.00000

Variables

Starting Values						Final Results			
No.	Variable Name	Status	Initial Value	Lower Bound	Upper Bound	No.	Variable Name	Initial Value	Final Value
1	X	UL	1.00000	0	1	1	X	1.00000	1.00000
2	X	UL	1.00000	0	1	2	X	1.00000	0.00000
3	X	UL	1.00000	0	1	3	X	1.00000	0.00000
4	X	UL	1.00000	0	1	4	X	1.00000	1.00000
5	X	UL	1.00000	0	1	5	X	1.00000	0.00000
6	X	UL	1.00000	0	1	6	X	1.00000	0.00000
7	X	UL	1.00000	0	1	7	X	1.00000	0.00000
8	X	UL	1.00000	0	1	8	X	1.00000	0.00000
9	X	UL	1.00000	0	1	9	X	1.00000	0.00000
10	X	UL	1.00000	0	1	10	X	1.00000	0.00000
11	X	UL	1.00000	0	1	11	X	1.00000	1.00000
12	X	UL	1.00000	0	1	12	X	1.00000	1.00000

No.	Objective Function	Binding Constrs	Super Basics	Infeas Constr	Norm of Red. Grad	Hessian Cond. No.	Step Size	Degen Step
1	3205.43710	0	12	2	0.57590	1	0	
2	3.55285	0	11	1	0.28146	1	1	
3	2.88211	0	10	1	0.34697	1	0.061	



Optimization is used to allocate resources where the results provide the max returns or the min cost/risks. Uses include managing inventories, financial portfolio allocation, product mix, project selection, etc.

Objective

Method

Constraints

Statistics

Decision Variables

☐

Static Optimization
 Run on static model without simulations. Usually run to determine the initial optimal portfolio before more advanced optimizations are applied.

☐

Dynamic Optimization
 A simulation is first run, the results of the simulation are applied in the model, and then an optimization is applied to the simulated values.

 Number of Simulation Trials

☒

Stochastic Optimization
 Similar to dynamic optimization but the process is repeated several times. The final decision variables will each have its own forecast chart indicating its optimal range.

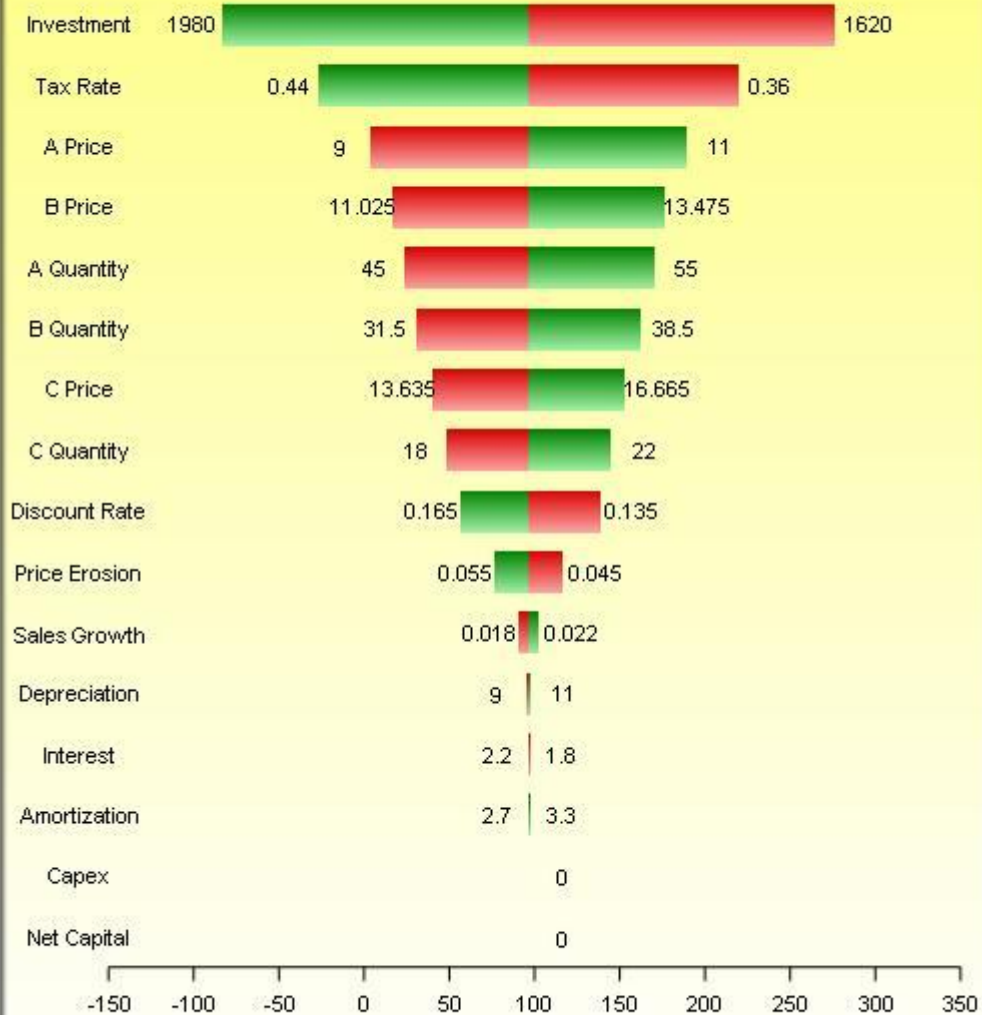
 Number of Simulation Trials
 Number of Optimization Runs

Advanced

OK

Cancel

Tornado Chart



<i>Normal (Single Fit)</i>
93.75
99.66
86.95
111.86
99.55
95.55
97.32
87.25
90.68
85.86
98.74
88.76
97.70
99.75
90.05
106.63
103.21
66.48
104.38
123.26
103.65
92.85
84.18
109.85
86.04
102.26
105.36
97.64
109.15
110.98
108.09
95.38
93.21

<i>Student's T</i>	<i>Triangular</i>	<i>Uniform</i>
47.56	185.86	53.30
49.71	204.77	53.09
50.24	145.61	52.09
50.36	219.85	45.81

Single-Fit

Distribution fitting takes existing raw data and statistically finds the best-fitting distribution (i.e., by optimizing the parameters of each distribution and performing statistical hypotheses tests).

Distribution Type

☒ Fit to Continuous Distributions
 ☐ Fit to Discrete Distributions

Select Distributions to Fit:

☒

Cauchy Distribution

☒

ChiSquare Distribution

☒

Exponential Distribution

☒

F Distribution

☒

Gamma Distribution

☒

Gumbel (Maximum) Distribution

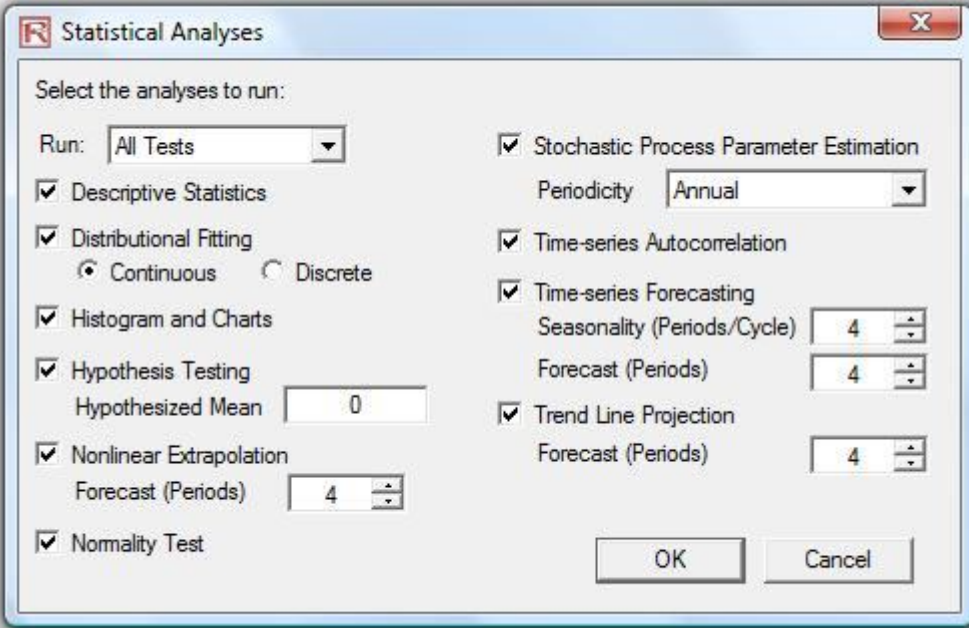
Select All

Clear All

OK

Cancel

52.25	128.85	49.08
49.01	166.19	52.81
50.51	197.52	50.74
49.72	279.06	47.98

A screenshot of a software dialog box titled "Statistical Analyses". The dialog box has a standard Windows-style title bar with a close button (X) in the top right corner. The main area is divided into two columns of options. The left column contains a "Run:" dropdown menu set to "All Tests", followed by a list of analysis categories, each with a checked checkbox: "Descriptive Statistics", "Distributional Fitting" (with sub-options "Continuous" selected and "Discrete" unselected), "Histogram and Charts", "Hypothesis Testing" (with a "Hypothesized Mean" input field containing "0"), "Nonlinear Extrapolation" (with a "Forecast (Periods)" spinner set to "4"), and "Normality Test". The right column contains a list of specific analysis methods, each with a checked checkbox: "Stochastic Process Parameter Estimation" (with a "Periodicity" dropdown set to "Annual"), "Time-series Autocorrelation", "Time-series Forecasting" (with "Seasonality (Periods/Cycle)" and "Forecast (Periods)" spinners both set to "4"), and "Trend Line Projection" (with a "Forecast (Periods)" spinner set to "4"). At the bottom right, there are "OK" and "Cancel" buttons.

Statistical Analyses

Select the analyses to run:

Run: All Tests

☒ Descriptive Statistics

☒ Distributional Fitting

☒ Continuous ☐ Discrete

☒ Histogram and Charts

☒ Hypothesis Testing

 Hypothesized Mean: 0

☒ Nonlinear Extrapolation

 Forecast (Periods): 4

☒ Normality Test

☒ Stochastic Process Parameter Estimation

 Periodicity: Annual

☒ Time-series Autocorrelation

☒ Time-series Forecasting

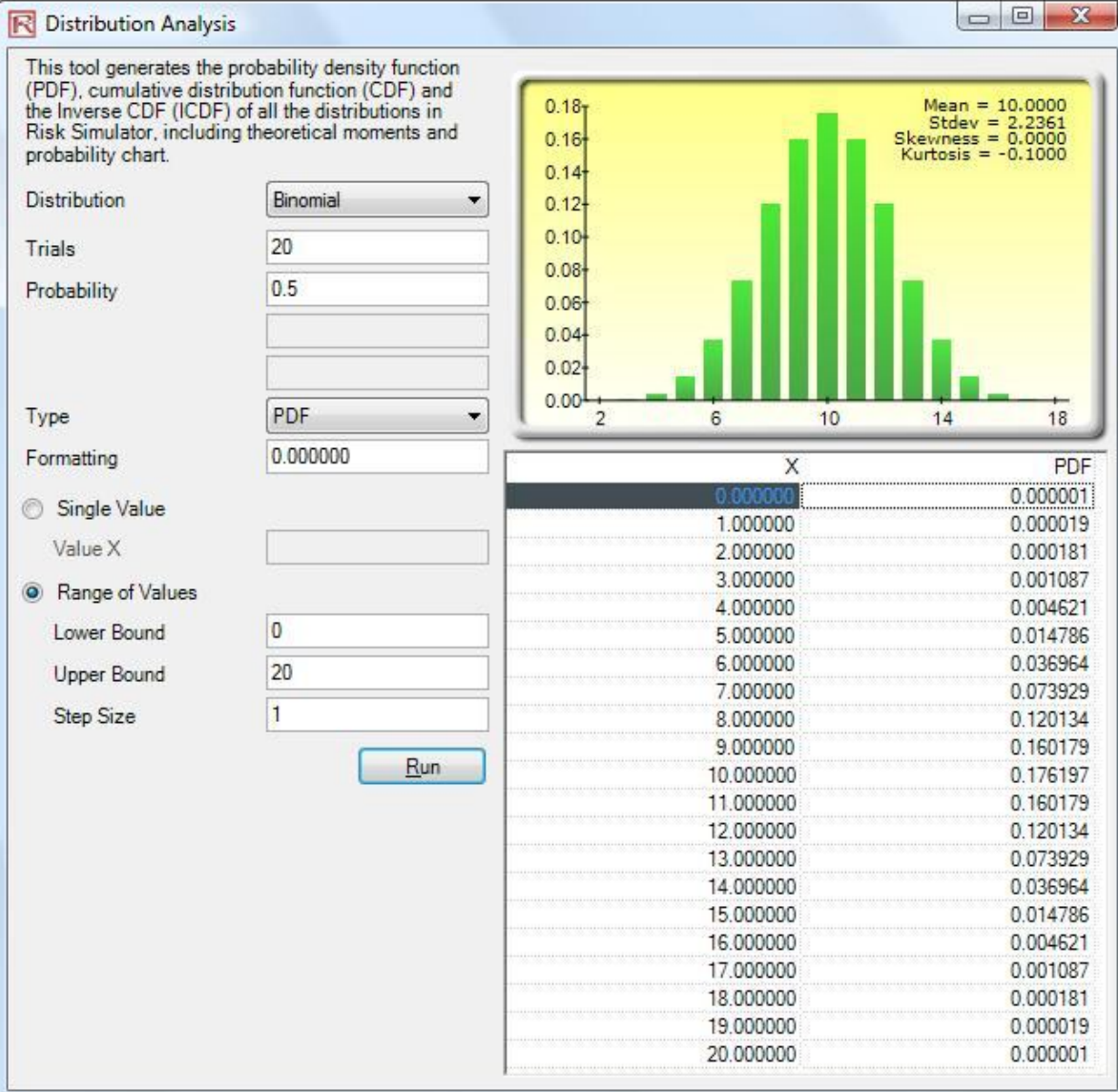
 Seasonality (Periods/Cycle): 4

 Forecast (Periods): 4

☒ Trend Line Projection

 Forecast (Periods): 4

OK Cancel



Options

☐ Minimize Excel and All Charts When Running

☒ Start Risk Simulator with Excel

☐ Always Show Forecast Windows on Top

☐ Show Cell Comments on Assumptions, Forecasts and Decision Variables

Correlation

☒ Normal Copula (Default)

☐ T Copula: DF =

☐ Quasi-Normal Copula: DF =

Parameters' Color Scheme

Assumption

Decision

Forecast

Language

English

* LHS is not recommended when there are correlated assumptions

OK

Cancel

Random Number Generator

☒ ROV Risk Simulator (Default)

☐ Advanced Subtractive Random Shuffle

☐ Long Period Shuffle

☐ Portable Random Shuffle

☐ Quick IEEE Hex

☐ Basic Minimal Portable

Simulation

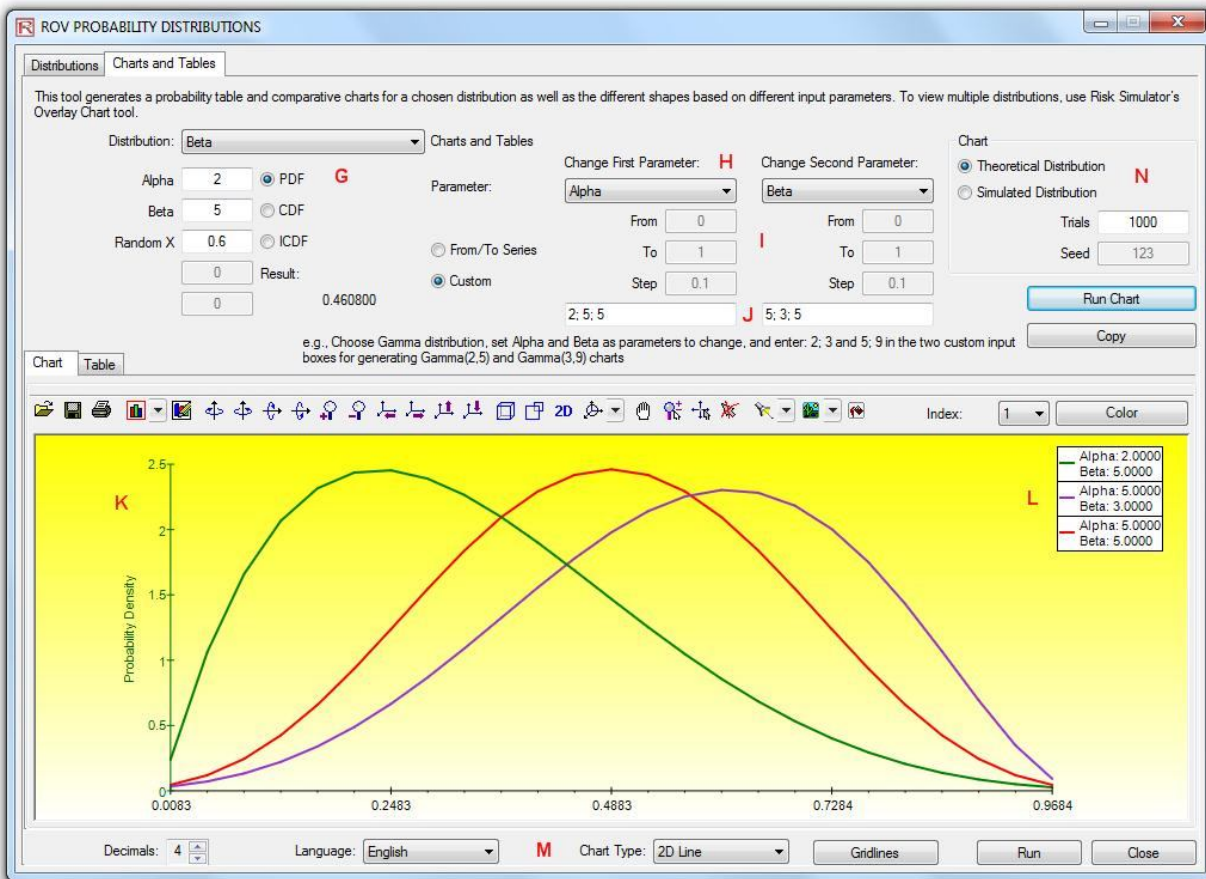
☒ Monte Carlo Simulation (Default)

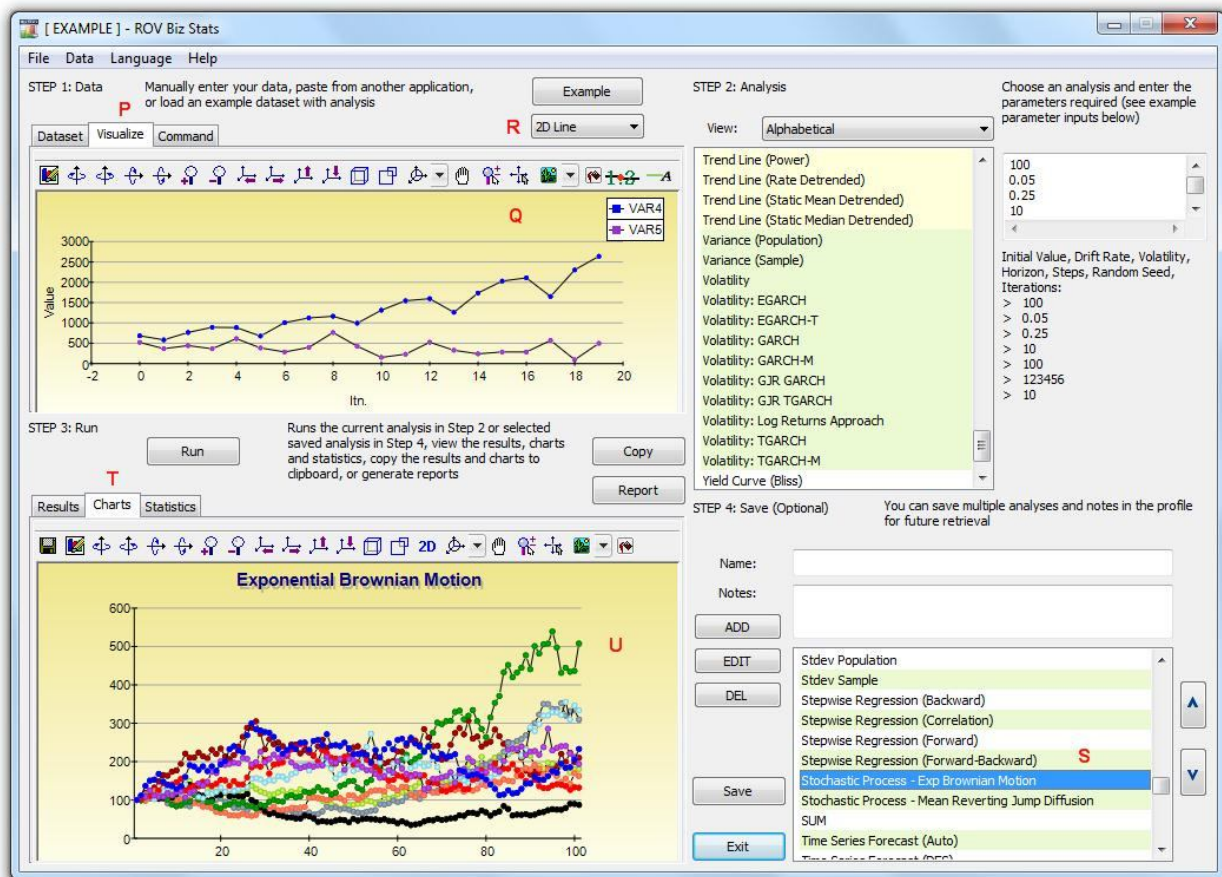
☐ Latin Hypercube Sampling (LHS)*

Low Sampling (Faster)

Balanced

High Sampling (Slower)





Neural Network Forecast

STEP 1: Data

Manually enter your data, paste from another application, or load an example dataset with analysis

Paste

N	VAR2	VAR3	VAR4	VAR5	VAR6	VAR7	VAR8	VAR9	VAR10	VAR11
NOT...										
1	1	459.11								
2	2	460.71								
3	3	460.34								
4	4	460.68								
5	5	460.83								
6	6	461.68								
7	7	461.66								
8	8	461.64								
9	9	465.97								
10	10	469.38								

STEP 2: Choose analysis type, variable, and forecast period to run:

English

VAR2

3

210

210

Copy

Run

☐ Cosine with Hyperbolic Tangent
 ☒ Hyperbolic Tangent
 ☐ Linear
 ☐ Logistic

Layers:
 Testing Periods:
 Forecast Periods:

☒ Apply Multiphased Optimization

Results

Charts

Sum of Squared Errors (Training) : 1.822044
RMSE (Training) : 0.093820
Sum of Squared Errors (Modified) : 59375.218349
RMSE (Modified) : 16.814849
Forecasting
* indicates negative values

Period	Actual (Y)	Forecast (F)	Error (E)
211	581.5000	613.3528	*31.8528
212	584.2200	613.5197	*29.2997
213	589.7200	613.6203	*23.9003
214	590.5700	613.7188	*23.1488
215	588.4600	613.8520	*25.3920
216	586.3200	614.0608	*27.7408
217	591.7100	614.2046	*22.4946
218	593.2600	614.3029	*21.0429
219	592.7200	614.4223	*21.7023
220	592.3000	614.5671	*22.2671
221	589.2900	614.7154	*25.4254
222	593.9600	614.8963	*20.9363
223	597.3400	614.9954	*17.6554
224	600.0700	615.0992	*15.0292
225	596.8500	615.2115	*18.3615

Combinatorial Fuzzy Logic Forecast

STEP 1: Data

Manually enter your data, paste from another application, or load an example dataset with analysis

Paste

N	VAR1	VAR2	VAR3	VAR4	VAR5	VAR6	VAR7	VAR8	VAR9	VAR10
NOT...	FUZZY									
1	684.20									
2	584.10									
3	765.40									
4	892.30									
5	885.40									
6	677.00									
7	1006.60									
8	1122.10									
9	1163.40									
10	993.20									

STEP 2: Enter required inputs and select the variable to forecast

English

VAR1

Seasonality: 4

Forecast Periods: 10

Copy

Run

Results

Charts

Results RMSE : 707.039492

Auto ARIMA RMSE : 249.495091

Time-Series Auto RMSE : 287.252763

Trend Line Exponential RMSE : 775.403678

Trend Line Linear RMSE : 912.616213

Trend Line Logarithmic RMSE : 1488.012692

Trend Line Moving Average RMSE : 988.333906

Trend Line Polynomial RMSE : 758.307610

Trend Line Power RMSE : 1268.660480

RESULTS

Forecast Fit

* indicates negative values

Period	Actual (Y)	Forecast (F)	Error (E)
1	684.2000		
2	584.1000		
3	765.4000		
4	892.3000		
5	885.4000	802.4484	82.9516
6	677.0000	863.9179	*186.9179
7	1006.6000	971.7020	34.8980
8	1122.1000	1083.6028	38.4972

Genetic Algorithm

Genetic

Objective Cell:  ☒ Maximize ☐ Minimize

Variables:

ColumnCell	ColumnMin	ColumnMax
------------	-----------	-----------

Constraints:

ColumnCell	ColumnMin	ColumnMax
------------	-----------	-----------

Max Iterations:

Mutation Rate:

Population Size:

Diversity:

Crossover Rate:

Elitism:

Cross over:

Unchange:

Result: