

# FasterAnalytics for Financial Services – A Case Study

## Introduction

DecisionQ has developed FasterAnalytics, a unique analytics package that enables analysts and managers to use sophisticated predictive analytics from their desktops. FasterAnalytics is fast and creates high quality, predictive models from data that enable day-to-day review of financial and trading data, real-time hypothesis testing, and rapid decisions.

FasterAnalytics uses a modeling approach called Bayesian Networks to provide a mapping of the complex relationships in data, which can then be used to make high quality predictions. Users can:

- Get an instant global view of their data.
- Understand the driving factors in the data.
- Test hypotheses in real time in our model Explorer.
- Produce reports that can be exported to other applications.
- Make determinations that can help prioritize the use of scarce research resources.

#### Market Overview

The Financial Services industry spends billions of dollars on research globally. It is served by an array of software vendors selling tools that enable data analysis and modeling. The pace of financial markets requires that financial professionals constantly revise their models to minimize risk while maximizing profit.

#### Value to the Customer

FasterAnalytics enables both experts and non-experts in statistics to discover and leverage knowledge from large quantities of data quickly. Examples include:

- Automatically mapping data where targets are unknown to reveal correlations.
- Identifying both positively and negatively correlated relationships to a target variable.
- Discovering new relationships between variables and identifying new profit opportunities.
- Predicting the behavior of any factor or combination of factors in the model.
- Allowing analysts to develop new models in minutes, keeping pace with shifting markets.

FasterAnalytics is designed for real-time environments. Bayesian models are highly effective at identifying emerging trends that can be used to either to identify emerging profit opportunities or alert financial analysts to potential risk exposure.

### Product and Technology

DecisionQ Corporation has produced a range of modules that perform data analysis, modeling, visualization, reporting, and decision optimization. FasterAnalytics modules include:

- *Discretizer.* Automatically configures the data for modeling.
- *Modeler.* Quickly creates a visual model of the data.
- *Explorer.* Allows real-time generation and testing of hypotheses.
- *Reporter.* Extracts insights and key points for inclusion in reports and presentations.

# Using the System: An Arbitrage Example

The following is an example application of our software using publicly available market data. We have used a set comprised of 5 years of closing data across 42 indices, derivatives, and commodities. FasterAnalytics built the model in this example, from start to finish, in less than 15 minutes.

To build predictive models, our learning engine requires the data to be in a flat tabular format. (Figure 1) The data can be numerical, or variable character strings. Our software also handles missing values automatically and will either impute a value or treat missing values as a special category, at the user's discretion.

# Figure 1: This example uses a financial data set held in an Excel spreadsheet as shown below (Partial).

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Having selected the data, a fully automated process will continue until a full model is presented, or the user can stop each part of the process to manually change parameters in order to leverage particular domain expertise. The software begins by categorizing the data and 'binning' in accordance with the default settings; the data is then passed seamlessly to the Modeler for automated model development. Once the software has mapped all the complex correlations and causality in the data a graphical model is presented in the Explorer (Figure 2). This whole process takes only minutes.



Figure 2: Base case model of the data presented in Explorer

The display illustrates conditional dependence between variables and the pathways existing in the trading model. Notice that the network has several branches, one consisting of interest rates and commodity prices, one consisting of trading volume and directional data, and one consisting of index data.

In the example below, we examine how index performance affects volatility in bull vs. bear markets. We begin by selecting our target variable, CBOE NASDAQ VOLATILITY INDEX. The thick border indicates that this is the target selected, and its color red indicates that we are interested in analyzing how other variables behave when the target is 'high'. The coloring of the remaining nodes is red if the corresponding variable values are also 'high', and green if the corresponding variable values are also 'high' measure of expression level.



Figure 3: CBOE NASDAQ VOLATILITY INDEX set to "high"

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Figure 4: The complementary case analyzing CBOE NASDAQ VOLATILITY INDEX set to "low"



Compare the two models in Figure 3 and 4 above with the base level in Figure 1. It is also possible to select two or more variables simultaneously. The extent to which CBOE NASDAQ VOLATILITY INDEX is related to other variables in its neighborhood is intuitive and clear. This can be used to search for drivers of NASDAQ volatility.

Each variable can be expanded using the 'View' menu or icons to show quantitative information about the relationships. The population data is displayed as "cases" with bars that represent the marginal probability distribution of each case.

Suppose that we are interested in examining how NASDAQ DECLINES and NASDAQ ADVANCES affect CBOE NASDAQ VOLATILITY. We first select these nodes and click "Graph" to display the cases within these nodes. (Figure 5) This can be done for as many variables as we may choose.

Figure 5: Quantitative information about NASDAQ declines and advances relative to volatility.



If we wish to test hypotheses, we can modify any node and see how our hypothesis affects the model. Notice how information flows through the network.

When we change the state of declines or advances, we see very clearly that a declining market has a higher level of volatility than an advancing market. We can see this graphically in Figures 6a and 6b below.



Figure 6a: NASDAQ advances and their effects on volatility.





The Reporter module can be used to create a report that will show the conditional probabilities (or predicted likelihood) of any target variables, given the expression of any independent variable(s). Any part of the model visualization can be pasted into the Reporter module and then transferred into other applications. Figure 7 shows a sample report.

Figure 7: A sample report listing the probabilities of volatility given advances and declines.

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DecisionQ sells predictive modeling software and complementary professional services. Alternatively, components from FasterAnalytics can be integrated into third party applications as part of broad data management and analysis platform.

If you have any further questions or would like to schedule a more detailed demonstration in person or over the web, please contact us.

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