Unleash the Power of Predictive Analytics with the SAP HANA® Platform
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Big Data, Big Challenges, and Big Opportunities

Understanding what happened in the past can provide significant insights, but to truly benefit from your application and data infrastructure investments, you want to understand how to make better decisions to improve your future. Simply comprehending how to avoid an unexpected turn of events could save a lot of money, just as knowing what you could do to increase revenues can make a huge difference to your bottom line.

Traditional business intelligence (BI) tools provide limited explanations of why something happened, because most BI solutions are geared more for reporting and dashboarding workflows that focus on the past. As a result, users are left to create their own insights from slicing, dicing, and drilling through the data themselves.

The left side of Figure 1 depicts the evolution of analytics from a pure presentation of raw data to today’s concept of descriptive analytics that helps users understand what their data says about the past. The right side of the figure shows the natural evolution of BI toward predictive analytics. Predictive analytics uses the understanding of the factors influencing the past to determine what could happen next and even what could be done to positively influence future outcomes.

Predictive analytics algorithmically analyzes historical data to find trends and patterns to understand relationships between elements. By understanding these relationships, you can make better decisions. For example, knowing a particular product generates more profit per unit versus other products may lead you to create a special promotion to increase sales.

Creating such predictive insight is key in Big Data scenarios where a single data row by itself may not have special significance, but its contribution to an overall trend, pattern, or calculation could be meaningful. These relationships are difficult to uncover through traditional BI techniques because they may be too subtle or too complicated to be visualized on a chart. Mathematical analysis using statistical techniques has proved to be significantly more effective at understanding complex patterns in very large volumes of disaggregated data.
Figure 1: Moving Decision Making from Sense and Respond to Predict and Act

Legend
EDW = enterprise data warehouse
ETL = extract, transact, load
BI = business intelligence
OLAP = online analytical processing
The result of predictive analytics is not a BI artifact like a query or a report, but a mathematical representation (a “model”) of the relationships found between different elements in the data. You can then use this model to perform what-if analysis to maximize or minimize a value. Or you can apply the model against entirely new data to predict a future outcome such as “a customer’s probability to purchase” or “projected revenue for the next four quarters.”

Making predictions in real time is even more valuable: rather than waiting until the end of the day or end of the week to run an analysis, you can make a decision as soon as data becomes available. This can mean millions of dollars or, in some industries, could be the difference between life and death.

However, most organizations have many heterogeneous technologies supporting legacy environments, leading to inefficient data management and significant operational overhead. In addition, many organizational silos are common in larger organizations where multiple teams are responsible for the data infrastructure. This is illustrated in Figure 2.

There is an ever-increasing set of technologies designed to make each of these steps faster. However, both the amount of data being generated and the speed at which this data needs to be analyzed is growing exponentially so that traditional methods of analysis will soon no longer be viable.

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**Figure 2: Data Analysis – Multiple Silos Versus In-Memory Computing**

![Figure 2: Data Analysis – Multiple Silos Versus In-Memory Computing](image)

**Separated transactions + analysis + acceleration processes**
- Redundant data in and across applications
- Inherent data latency

**One in-memory atomic copy of data for transactions + analysis**
- Elimination of unnecessary complexity and latency
- Acceleration through simplification

**Legend**
ETL = extract, transact, load
SAP HANA lets you move from descriptive analytics to predictive analytics while taking advantage of the speed and scalability provided by a native in-memory solution. SAP® BusinessObjects™ Predictive Analytics software with SAP HANA further improves this value. It makes fully automated predictive capabilities accessible to a wider range of users without requiring them to have data science experience or a statistical background. This is accomplished through the automated predictive library (APL) for SAP HANA, which is a native implementation of our mature, proprietary, automated machine learning technology that can handle large volumes of data at lightning speed.

The industry is turning to in-memory technologies that not only store data within volatile memory but also perform computations on that data in place. Such a strategy brings the calculations to the data rather than the other way around. This avoids much of the data extraction, transfer, and duplication required in heterogeneous analytical environments. Using sophisticated onboard calculation and prediction engines, in-memory platforms such as SAP HANA® are changing how we process, analyze, and act on the ever-increasing volume of data we create every day (see Figure 3).

Figure 3: Using In-Memory Technology to Process, Analyze, and Act on Big Data

<table>
<thead>
<tr>
<th>What happened?</th>
<th>Why did it happen?</th>
<th>What will happen?</th>
<th>How can I make it happen?</th>
</tr>
</thead>
<tbody>
<tr>
<td>• Standard reports</td>
<td>• Drill-down analysis</td>
<td>Predictive modeling</td>
<td>Real-time automated predictive analysis</td>
</tr>
<tr>
<td>• Ad hoc reports</td>
<td>• Data discovery</td>
<td></td>
<td></td>
</tr>
<tr>
<td>• OLAP analysis</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Sense and respond

Predict and act

SAP HANA®

Legend

OLAP = online analytical processing
Examining SAP BusinessObjects Predictive Analytics

SAP BusinessObjects Predictive Analytics is a comprehensive predictive solution that consists of a desktop application and a number of server-based technologies that can be deployed either on SAP HANA or in configurations with other platforms. In this document, we focus on SAP BusinessObjects Predictive Analytics when used in SAP HANA platform environments, but many of the principles apply to deployments on other platforms as well.

Providing statistical analysis and data-mining functionality, SAP BusinessObjects Predictive Analytics lets users perform a variety of predictive workflows locally on the desktop and then work in a server environment to manage the entire predictive lifecycle. Users from business analysts to more-seasoned data scientists can build predictive models, discover hidden relationships in their data, and then use these insights to better understand historical data or even make predictions about future events.

THE DESKTOP
SAP BusinessObjects Predictive Analytics provides a stand-alone desktop application that serves as the recommended front end for all predictive technologies from SAP. This tool is targeted primarily at business and data analysts to easily find answers to their business questions without requiring extensive training or data science skills. However, it also has an “expert” mode for those who desire a more hands-on predictive experience.

At the core of the solution are proprietary automated machine learning algorithms designed to make predictive technologies more accessible and easier to use. Using a wizard-driven interface, users are guided through the predictive modeling process of data preparation, model creation, and model deployment (see Figure 4).

Figure 4: Main Interface of SAP® BusinessObjects™ Predictive Analytics
MAJOR FEATURES OF SAP® BUSINESSOBJECTS™ PREDICTIVE ANALYTICS

Prepare data:
✓ Define source data creation or manipulations
✓ Create persistent metadata definitions
✓ Enable the automatic creation of thousands of derived attributes

Perform automatic predictive modeling:
✓ Regression and classification
✓ Clustering
✓ Forecasting
✓ Association rules
✓ Social network analysis

Produce automatic analysis of models:
✓ Estimate predictive power and confidence
✓ Identify the most contributive attributes
✓ Produce statistical reports
✓ Create graphical analysis of results

Save, export, and apply results:
✓ Create and save persistent models
✓ Export scoring equations as structured query language (SQL) or code
✓ Apply predictive models directly to data sets and in your database (if applicable)
If you are already using SAP HANA, you can use the platform’s database tables and analytical and calculation views as sources for your predictive models.

For more control in the predictive-modeling process, you can use SAP BusinessObjects Predictive Analytics as a graphical workbench with a flow-based paradigm. You can graphically drag and drop algorithms, chain them together, and perform automated model comparison from the same screen. In this mode, you can use native libraries in SAP BusinessObjects Predictive Analytics, including the predictive analysis library (PAL) for SAP HANA, the automated predictive library (APL) for SAP HANA, and any algorithm created using the open-source R language, as seen in Figure 5.

**Figure 5: Flow-Based Paradigm Using R Language**
INTEGRATION WITH SAP BUSINESS WAREHOUSE
You may be using the SAP Business Warehouse (SAP BW) application or SAP BW powered by SAP HANA for your warehousing needs. If so, you can use the expert interface to connect SAP BusinessObjects Predictive Analytics directly to SAP BW or SAP BW on SAP HANA using a native BI consumer services (BICS) connection. By connecting to info providers of SAP BW and executing SAP Business Explorer® (SAP BEx) tools queries, you can download data sets for offline model training (see Figure 6). You can even eliminate the need to manually export data to a file for predictive analysis.

Figure 6: SAP BusinessObjects Predictive Analytics Connecting to SAP Business Warehouse Using BICS

CLIENT-SERVER PREDICTIVE ANALYTICS
You can deploy SAP BusinessObjects Predictive Analytics in a client-server configuration to create and process models remotely and publish them to a central repository for sharing with other users. The addition of a predictive analytics server in the environment enables you to move from executing a single model interactively to automating the creation, execution, and validation of thousands of models. And you can do this without needing to be in front of your desktop.

The user component of this server is model manager, a thin-client, Web-based application that allows all users to automate as much of the predictive lifecycle as possible, such as:
• Retraining a model
• Applying a model to a new data set
• Detecting model deviations
• Detecting deviation of a data set

In a database environment that is not using SAP HANA, the predictive analytics server extracts data from the source database and computes the models locally in the same manner as the desktop to perform its functions.

Note: When using a license for SAP BusinessObjects Predictive Analytics software for use with databases other than SAP HANA, you cannot use the software with SAP HANA without obtaining the corresponding license. In addition, you cannot use the desktop of SAP BusinessObjects Predictive Analytics with SAP HANA–specific features such as the predictive analysis library or integration to an external R server.
USING SAP BUSINESSOBJECTS PREDICTIVE ANALYTICS AND SAP HANA

SAP BusinessObjects Predictive Analytics with SAP HANA provides the same capabilities as described in the previous section and much more. SAP BusinessObjects Predictive Analytics is optimized for SAP HANA to support huge data volumes and in-memory processing. You can connect to SAP HANA in an “online mode” that pushes calculation operations to the APL and PAL for SAP HANA. This means you do not need to extract the data and perform those same calculations on the desktop. In addition, there is no data movement required for SAP HANA–based predictive workflows, and only the predictive results are sent back to the desktop client after processing.

SAP BusinessObjects Predictive Analytics supports the full range of predictive capabilities in SAP HANA, including creating multiple models using any combination of PAL, APL, and R server scripts. The solution also supports executing advanced functions from SAP HANA, such as unified demand forecast (UDF), sentiment analysis, and optimization function libraries (OFL).

In an SAP HANA platform environment, all model management functions in SAP BusinessObjects Predictive Analytics are performed with the predictive analytics server. The ability to manage the models directly from an interface on SAP HANA is part of the longer-term road map.

Note: When using a license for SAP BusinessObjects Predictive Analytics with SAP HANA, you will not be able to use the automated analytics server with other databases without additional licensing.

Using sophisticated onboard calculation and prediction engines, in-memory platforms such as SAP HANA are changing how we process, analyze, and act on the ever-increasing volume of data we create every day.
SAP HANA As a Predictive Analytics Platform

SAP HANA was introduced in 2010 as a truly “in-memory” database and computational platform. Its ability to implement execution engines that can operate on data in-memory enables it to offer multiple predictive capabilities and offer them concurrently. The platform has steadily added newer native onboard capabilities such as text analysis, geospatial computation, and predictive analysis.

Today, a single data platform can support the different predictive needs of business analysts, data scientists, and even application developers in an organization. The predictive engines of SAP HANA are implemented at the application function library (AFL) layer, which enables almost any other process to use or embed these predictive capabilities with minimal effort (see Figure 7).

Figure 7: Logical View of SAP BusinessObjects Predictive Analytics with SAP HANA

Legend
API = application programming interface
LoB = line of business
CRM = customer relationship management
ERP = enterprise resource planning
The following sections discuss many of the benefits and considerations of using SAP BusinessObjects Predictive Analytics with SAP HANA as your data processing and analytical platform.

**SPEED AND EFFICIENCY**

The SAP HANA platform’s unique ability to use native, onboard execution engines means you can perform predictive calculations for training, validating, and scoring without data extraction or even manipulation. Running predictive models in-memory results in dramatic performance improvements in addition to a more-efficient use of system resources. If you can make faster decisions based on real-time analysis of data, you can typically make better and more-informed decisions. When these execution engines are combined with the event-processing capabilities of the platform, you can be instantly alerted and take action as soon as a situation or opportunity is detected.

Consider a scenario where you are using millions of customer records in an analysis to determine clusters of attractive market segments within the population. Processing such a large number of records is already an expensive and time-consuming task. However, extracting and transferring these records to a separate analytics server also creates additional workload and cost for the rest of the IT infrastructure.

The ability of SAP HANA to perform predictive calculations directly from within SQLScript enables you to use existing SQL expertise in your organization while providing on-the-fly processing on compressed data with unmatched speed and efficiency.

**EASE OF USE**

The desktop of SAP BusinessObjects Predictive Analytics is recommended for all deployments of SAP HANA because it provides a GUI for all predictive technologies in SAP HANA such as the PAL and APL. The software includes a GUI for third-party technologies as well, such as scripts using the open-source R language. You can use any combination of these technologies together while taking advantage of server-side processing and native support of database tables and calculation and analytical views of SAP HANA.

Providing all business analysts and data scientists with the desktop of SAP BusinessObjects Predictive Analytics helps ensure that they can access data and native predictive technologies of SAP HANA without having to acquire new skills. This includes not having to learn SQLScript of SAP HANA or use a more administratively focused tool such as the SAP HANA studio. This dramatically increases the number of people who can access predictive capabilities as well as the potential return on your investment in SAP HANA.
SIMPLICITY AND AUTOMATION
In most business scenarios, speed is not enough; if the power to achieve the benefits of predictive analytics at such speeds is limited to a handful of “experts,” the number of scenarios to which predictive analytics can bring real business value will also be limited. The reality is that most organizations have very few people with data science backgrounds and have even fewer data scientists. The traditional approach to solving this problem is to train more people and create an even greater investment in data science skills.

The automated predictive library for SAP HANA implements a sophisticated automated machine learning engine to provide a “data-scientist-in-a-box” solution that can generate complex predictive models on its own without requiring the user to have deep data science knowledge. We will cover the APL in more detail in the section “Automated Predictive Library: APL.”

The model manager tool within SAP BusinessObjects Predictive Analytics provides automatic performance-tuning capabilities to help ensure that models are tuned to be operating at peak performance at all times for optimal outcomes. The tool features a browser-based, single-sign-on environment and user-friendly scheduling interface designed for data analysts. As a result, you can try a variety of scenarios, incorporate the incremental improvements in the model in real time, schedule model refreshes, manage models by exception, and deploy scores instantly.

When combined with other components from SAP BusinessObjects Predictive Analytics, SAP HANA becomes a single platform for all predictive workflows and automates the full predictive lifecycle from model creation to deployment and even ongoing model validation. The IT landscape becomes simpler, and users have a single analytics platform regardless of the predictive technology they use.

OPERATIONALIZATION
When you operationalize a model, you use the predictive insights to improve a process, workflow, or decision in a production environment. This is an often-overlooked aspect of predictive analytics; creating a model in a desktop tool may provide an interesting analysis but does not provide much incremental business value.

You can maximize the return on your predictive investment by operationalizing your models and embedding them directly into your business processes. However, in traditional solutions, deploying a model is typically a manual step and may even require an additional predictive server to be configured.
Predictive models executed in SAP HANA provide “in-database” scoring on the fly and give you the flexibility to embed the results directly wherever they will be most useful. These include embedding into BI workflows, databases, business processes, or line-of-business solutions. The downstream users of these results don’t need to use the predictive models directly and may not even know they exist.

To illustrate, let’s take a classification example to determine insurance claims that are potentially fraudulent. Applying a predictive model to customer data results in a data set that includes additional fields with the results of the predictive calculations.

Figure 8 shows the results of a classification model, based on APL, to determine fraudulent insurance claims. In this example, claims with a fraud score of more than 60% are more likely to be the more-fraudulent ones.

You can store these results in a table or view of SAP HANA and repopulate them in real time for subsequent queries.

<table>
<thead>
<tr>
<th>Claim ID</th>
<th>Prediction</th>
<th>Percent Likelihood</th>
<th>Fraud Score</th>
</tr>
</thead>
<tbody>
<tr>
<td>CL_1031721</td>
<td>Fraudulent Claim</td>
<td>55.55</td>
<td>86.38</td>
</tr>
<tr>
<td>CL_0995635</td>
<td>Fraudulent Claim</td>
<td>55.55</td>
<td>79.21</td>
</tr>
<tr>
<td>CL_0960379</td>
<td>Fraudulent Claim</td>
<td>55.55</td>
<td>78.68</td>
</tr>
<tr>
<td>CL_0993311</td>
<td>Fraudulent Claim</td>
<td>55.55</td>
<td>70.36</td>
</tr>
<tr>
<td>CL_0962914</td>
<td>Fraudulent Claim</td>
<td>55.55</td>
<td>64.28</td>
</tr>
<tr>
<td>CL_0977425</td>
<td>Fraudulent Claim</td>
<td>55.55</td>
<td>64.01</td>
</tr>
<tr>
<td>CL_0993189</td>
<td>Fraudulent Claim</td>
<td>52.77</td>
<td>63.85</td>
</tr>
<tr>
<td>CL_1027905</td>
<td>Fraudulent Claim</td>
<td>52.66</td>
<td>63.5</td>
</tr>
<tr>
<td>CL_0997307</td>
<td>Fraudulent Claim</td>
<td>52.6</td>
<td>63.32</td>
</tr>
<tr>
<td>CL_0982255</td>
<td>Fraudulent Claim</td>
<td>52.17</td>
<td>61.99</td>
</tr>
<tr>
<td>CL_0970439</td>
<td>Fraudulent Claim</td>
<td>52.17</td>
<td>61.99</td>
</tr>
<tr>
<td>CL_0973639</td>
<td>Fraudulent Claim</td>
<td>51.57</td>
<td>60.11</td>
</tr>
<tr>
<td>CL_0983978</td>
<td>Fraudulent Claim</td>
<td>51.57</td>
<td>60.11</td>
</tr>
<tr>
<td>CL_0960294</td>
<td>Fraudulent Claim</td>
<td>51.11</td>
<td>58.68</td>
</tr>
<tr>
<td>CL_1011207</td>
<td>Fraudulent Claim</td>
<td>51.02</td>
<td>58.43</td>
</tr>
<tr>
<td>CL_0983297</td>
<td>Fraudulent Claim</td>
<td>33.55</td>
<td>53.45</td>
</tr>
<tr>
<td>CL_1003027</td>
<td>Legitimate Claim</td>
<td>100</td>
<td>-4.17</td>
</tr>
<tr>
<td>CL_0967006</td>
<td>Legitimate Claim</td>
<td>100</td>
<td>-4.51</td>
</tr>
</tbody>
</table>
You can consume this in-database scoring from within any other application; any user accessing this table will get the predictive scores, potentially without realizing that each record is the result of a predictive calculation. Furthermore, you can keep the model up-to-date automatically so the next refresh of the data set ensures that the most recent iteration of the predictive model is applied to the full result set.

Figure 9 takes this example one step further to illustrate the full end-to-end behavior.

In this example, the predictive model was created using SAP BusinessObjects Predictive Analytics and then published to SAP HANA. The APL is invoked using SQLScript to create a view from SAP HANA. This view contains the original data as well as an additional column that contains a predicted score and is consumed by SAP BusinessObjects Lumira software.

Using SAP BusinessObjects Lumira, you can then analyze the data, create visualizations and storyboards, and publish the results to a sharing platform, such as one of the server versions for SAP BusinessObjects Lumira. This enables you to share predictive results and the insights gained from SAP BusinessObjects Predictive Analytics with anyone in the organization and even externally.
REAL-TIME AUTOMATED PREDICTIVE ANALYSIS USE CASES
You can leverage the in-memory computation abilities of SAP HANA to satisfy real-time predictive analysis workloads without the added cost and overhead of a separate predictive solution. Here are a few example use cases that take advantage of SAP HANA platform capabilities:

<table>
<thead>
<tr>
<th><strong>Brand sentiment</strong></th>
<th>Analyze customer sentiment from social media sites such as Facebook, Twitter, and LinkedIn to improve customer experience and dynamically optimize market campaign performance.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Predictive maintenance</strong></td>
<td>Predict machinery performance degradation or potential equipment failure by continuously analyzing streams of machine data and diagnostic information.</td>
</tr>
<tr>
<td><strong>Insider threats</strong></td>
<td>Detect anomalies hidden in data about user behavior, and identify suspicious behavior to pinpoint potentially high-risk employees.</td>
</tr>
<tr>
<td><strong>Network optimization</strong></td>
<td>Understand usage patterns and predict customer trends to optimize network communications.</td>
</tr>
<tr>
<td><strong>Propensity to churn</strong></td>
<td>Determine a customer’s likelihood of turning over, and offer new services or support solutions in real time to proactively solve the customer’s problems and increase loyalty.</td>
</tr>
<tr>
<td><strong>Product recommendation</strong></td>
<td>Make highly targeted product recommendations by analyzing customer purchasing history and online browsing patterns to increase up-sell and cross-sell opportunities.</td>
</tr>
<tr>
<td><strong>Fraud detection</strong></td>
<td>Identify purchases or insurance claims that may have a high probability of being fraudulent by analyzing historical and transactional information in real time instead of after the fact.</td>
</tr>
<tr>
<td><strong>Real-time risk mitigation</strong></td>
<td>Identify high-risk events and proactively mitigate them by understanding the probability of their happening.</td>
</tr>
<tr>
<td><strong>360-degree customer view</strong></td>
<td>Build a more complete view of your customer by analyzing all data about a customer – including transactions, browsing history, customer profile, social media, and more – to understand behavior and motivations.</td>
</tr>
<tr>
<td><strong>Asset tracking</strong></td>
<td>Track high-value assets and identify abnormal behavior that may put assets at risk of loss, or identify inefficient usage that is costing your business money.</td>
</tr>
<tr>
<td><strong>Real-time demand-and-supply forecast</strong></td>
<td>Forecast customer demand by market, season, weather pattern, or market trend to predict supply requirements and optimize inventory.</td>
</tr>
</tbody>
</table>
If you use SAP BW powered by SAP HANA, SAP S/4HANA software, SAP HANA Live offerings, or virtually any application hosted on the SAP HANA platform, you can immediately use the onboard predictive capabilities directly. Or you can embed predictive directives directly into the application’s views from SAP HANA, or embed predictive directives directly into an application.

**ADDITIONAL BENEFITS OF SAP BUSINESSOBJECTS PREDICTIVE ANALYTICS WITH SAP HANA**

The APL and PAL capabilities for SAP HANA are implemented as application function libraries, which allows these engines to take advantage of everything the platform offers. This includes unified access to the information views and tables of SAP HANA, automatic parallelizing of all predictive computations, and full support of any application that uses SAP HANA.

SAP BusinessObjects Predictive Analytics is a stand-alone solution that can also leverage SAP HANA to support huge data volumes and in-memory processing without requiring data extraction.
“Algorithmic” Predictive Capabilities of SAP HANA: R and PAL

Data scientists typically create predictive models on historical data sets by applying one or more mathematical algorithms to capture the implicit relationships within the data. Use of these algorithms requires a significant understanding of the data and a solid grasp of statistics and other mathematical concepts (see Figure 10).

It is not uncommon for a data scientist to spend days or weeks analyzing the data before creating a robust and stable predictive model. In some cases, the data scientist may even have to create his or her own algorithms to solve more-complex problems or create a predictive model that is specific to an industry.

SAP HANA supports multiple methods for using data science–specific algorithms in the predictive process. The following section discusses the benefits and challenges of each.

R SCRIPTS
The open-source language R is the most popular predictive-modeling environment in the world. As a language created by mathematicians for mathematicians, it was designed from the beginning to be easily extensible and supports people sharing what they have written with others to use in their own modeling tasks. At the time of this writing, there are more than 5,800 open-source algorithms publicly available and countless others that are considered the proprietary intellectual property of their creators.

Figure 10: The Role of the Data Scientist

<table>
<thead>
<tr>
<th>THE WHAT</th>
<th>THE HOW</th>
</tr>
</thead>
<tbody>
<tr>
<td>Math and statistics</td>
<td>Process Statistical computing packages, such as R or PAL</td>
</tr>
<tr>
<td>• Statistical modeling</td>
<td></td>
</tr>
<tr>
<td>• Machine learning</td>
<td></td>
</tr>
</tbody>
</table>

Data scientist
SAP BusinessObjects Predictive Analytics offers an expert mode to desktop users to let you use any combination of algorithms from a variety of libraries. Libraries include those in SAP HANA such as the PAL and APL, as well as an external R server. This enables you to create predictive models on the desktop and then publish them to a server environment. There, you can execute the predictive workload without requiring any further intervention or involvement (see Figure 11).

Figure 11: Calling of R Scripts from SAP HANA
SAP HANA supports this type of sidecar deployment to offer data scientists the ultimate flexibility to use any algorithm they want or even create their own in R. They call R scripts through SQLScript in SAP HANA and then pass the scripts to an external R server, along with all of the required input data. The results are sent back to SAP HANA and combined with any local data required to complete the query (see Figure 12).

However, integrating the results from arbitrary scripting code using an external server results in significant overhead required to extract and transfer data to the R server for processing. Performing all calculations with a local in-memory predictive engine, such as the PAL for SAP HANA, avoids this overhead.

**PREDICTIVE ANALYSIS LIBRARY: PAL**

The predictive analysis library, or PAL, is designed to take advantage of the ability of SAP HANA to host execution engines and perform local calculations in memory. Unlike the previous option that uses an external predictive server for processing, this SAP HANA–native library enables users to perform in-database data mining and statistical calculations with excellent performance on large data sets.

While the PAL cannot replicate all 5,800 algorithms that are available with R, the PAL contains the SAP native C++ implementations of the most commonly used algorithms. The number of algorithms supported in PAL has been growing with every service pack of SAP HANA.
You can call PAL algorithms through an application such as SAP BusinessObjects Predictive Analytics or directly in SQLScript (see Figure 13).

Figure 13: Example SQLScript Using the PAL

```sql
/*The input, seed, parameter, and output tables must follow types specified in the signature table*/

POPULATE INPUT/OUTPUT TABLES
INSERT INTO PAL_AP_PDATA_TBL VALUES (1, 'DM_PAL', 'PAL_AP_DATA_T', 'IN');
INSERT INTO PAL_AP_PDATA_TBL VALUES (2, 'DM_PAL', 'PAL_AP_SEED_T', 'IN');
INSERT INTO PAL_AP_PDATA_TBL VALUES (3, 'DM_PAL', 'PAL_CONTROL_T', 'IN');
INSERT INTO PAL_AP_PDATA_TBL VALUES (4, 'DM_PAL', 'PAL_AP_RESULTS_T', 'OUT');

CALL AFL WRAPPER PROCEDURE TO CREATE THE PAL PROCEDURE
CALL SYS.AFLLANG_WRAPPER_PROCEDURE_CREATE('AFLPAL', 'AP', 'DM_PAL', 'PAL_AP',
PAL_AP_PDATA_TBL);

CREATE INPUT DATA TABLE
CREATE COLUMN TABLE PAL_AP_DATA_TBL LIKE PAL_AP_DATA_T;
INSERT INTO PAL_AP_DATA_TBL VALUES(1,0.10,0.10);
INSERT INTO PAL_AP_DATA_TBL VALUES(2,0.11,0.10);
INSERT INTO PAL_AP_DATA_TBL VALUES(3,0.10,0.11);
INSERT INTO PAL_AP_DATA_TBL VALUES(4,0.11,0.11);
INSERT INTO PAL_AP_DATA_TBL VALUES(5,0.12,0.11);

CREATE SEED DATA TABLE
CREATE COLUMN TABLE PAL_AP_SEED_TBL LIKE PAL_AP_SEED_T;

CREATE CONTROL TABLE
CREATE LOCAL TEMPORARY COLUMN TABLE #PAL_CONTROL_TBL LIKE PAL_CONTROL_T;

POPULATE CONTROL TABLE
INSERT INTO #PAL_CONTROL_TBL VALUES('THREAD_NUMBER',2,null,null);
INSERT INTO #PAL_CONTROL_TBL VALUES('MAX_ITERATION',500,null,null);
INSERT INTO #PAL_CONTROL_TBL VALUES('CON_ITERATION',100)null,null
INSERT INTO #PAL_CONTROL_TBL VALUES('DAMP',null,0.9,null);
INSERT INTO #PAL_CONTROL_TBL VALUES('PREFERENCE',null,0.5,null);
INSERT INTO #PAL_CONTROL_TBL VALUES('DISTANCE_METHOD',2,null,null);
INSERT INTO #PAL_CONTROL_TBL VALUES('CLUSTER_NUMBER',0,null,null);

CREATE RESULTS TABLE
CREATE COLUMN TABLE PAL_AP_RESULTS_TBL LIKE PAL_AP_RESULTS_T;

CALL PAL PROCEDURE
CALL DM_PAL.PAL_AP(PAL_AP_DATA_TBL, PAL_AP_SEED_TBL, #PAL_CONTROL_TBL, PAL_AP_RESULTS_TBL) with
OVERVIEW;
```
Figure 14 shows a nonexhaustive list of algorithms included in the PAL.

<table>
<thead>
<tr>
<th>Figure 14: Sample Algorithms in the Predictive Analysis Library</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Association Analysis</strong></td>
</tr>
<tr>
<td>• Apriori</td>
</tr>
<tr>
<td>• Apriori Lite</td>
</tr>
<tr>
<td>• FP-growth</td>
</tr>
<tr>
<td>• KORD – Top K Rule Discovery</td>
</tr>
<tr>
<td><strong>Regression</strong></td>
</tr>
<tr>
<td>• Multiple linear regression</td>
</tr>
<tr>
<td>• Polynomial regression</td>
</tr>
<tr>
<td>• Exponential regression</td>
</tr>
<tr>
<td>• Bivariate geometric regression</td>
</tr>
<tr>
<td>• Bivariate logarithmic regression</td>
</tr>
<tr>
<td><strong>Cluster Analysis</strong></td>
</tr>
<tr>
<td>• ABC Classification</td>
</tr>
<tr>
<td>• DBSCAN</td>
</tr>
<tr>
<td>• K-Means</td>
</tr>
<tr>
<td>• K-Medoid Clustering</td>
</tr>
<tr>
<td>• K-Medians</td>
</tr>
<tr>
<td>• Kohonen self-organizing maps</td>
</tr>
<tr>
<td>• Agglomerate hierarchical clustering</td>
</tr>
<tr>
<td>• Affinity propagation</td>
</tr>
<tr>
<td><strong>Classification Analysis</strong></td>
</tr>
<tr>
<td>• CART</td>
</tr>
<tr>
<td>• C4.5 decision tree analysis</td>
</tr>
<tr>
<td>• CHAID decision tree analysis</td>
</tr>
<tr>
<td>• K-nearest neighbor</td>
</tr>
<tr>
<td>• Logistic regression</td>
</tr>
<tr>
<td>• Back-propagation (neural network)</td>
</tr>
<tr>
<td>• Naive Bayes</td>
</tr>
<tr>
<td>• Support vector machine</td>
</tr>
<tr>
<td><strong>Time Series Analysis</strong></td>
</tr>
<tr>
<td>• Single exponential smoothing</td>
</tr>
<tr>
<td>• Double exponential smoothing</td>
</tr>
<tr>
<td>• Triple exponential smoothing</td>
</tr>
<tr>
<td>• Forecast smoothing</td>
</tr>
<tr>
<td>• Autoregressive integrated moving average (ARIMA)</td>
</tr>
<tr>
<td>• Brown exponential smoothing</td>
</tr>
<tr>
<td>• Croston method</td>
</tr>
<tr>
<td>• Forecast accuracy measure</td>
</tr>
<tr>
<td>• Linear regression with damped trend and seasonal adjust</td>
</tr>
<tr>
<td><strong>Probability Distribution</strong></td>
</tr>
<tr>
<td>• Distribution fit</td>
</tr>
<tr>
<td>• Cumulative distribution function</td>
</tr>
<tr>
<td>• Quantile function</td>
</tr>
<tr>
<td><strong>Outlier Detection</strong></td>
</tr>
<tr>
<td>• Interquartile range test (Tukey’s test)</td>
</tr>
<tr>
<td>• Variance test</td>
</tr>
<tr>
<td>• Anomaly detection</td>
</tr>
<tr>
<td><strong>Link Prediction</strong></td>
</tr>
<tr>
<td>• Common neighbors</td>
</tr>
<tr>
<td>• Jaccard coefficient</td>
</tr>
<tr>
<td>• Adamic/Adar</td>
</tr>
<tr>
<td>• Katzß</td>
</tr>
<tr>
<td><strong>Statistc Functions (Univariate)</strong></td>
</tr>
<tr>
<td>• Mean, median, variance, standard deviation</td>
</tr>
<tr>
<td>• Kurtosis</td>
</tr>
<tr>
<td>• Skewness</td>
</tr>
<tr>
<td><strong>Statistic Functions (Multivariate)</strong></td>
</tr>
<tr>
<td>• Covariance matrix</td>
</tr>
<tr>
<td>• Pearson correlations matrix</td>
</tr>
<tr>
<td>• Chi-squared tests:</td>
</tr>
<tr>
<td>− Test of quality of fit</td>
</tr>
<tr>
<td>− Test of independence</td>
</tr>
<tr>
<td>− F-test (variance equality test)</td>
</tr>
<tr>
<td><strong>Data Preparation</strong></td>
</tr>
<tr>
<td>• Sampling</td>
</tr>
<tr>
<td>• Random distribution sampling</td>
</tr>
<tr>
<td>• Binning</td>
</tr>
<tr>
<td>• Scaling</td>
</tr>
<tr>
<td>• Partitioning</td>
</tr>
<tr>
<td>• Principal component analysis (PCA)</td>
</tr>
<tr>
<td><strong>Other</strong></td>
</tr>
<tr>
<td>• Weighted scores table</td>
</tr>
<tr>
<td>• Substitute missing values</td>
</tr>
</tbody>
</table>
Note that data scientists are still free to use an external R server for algorithms or functions not included in the PAL. This enables any combination of algorithms to help ensure that users are benefiting from the speed of SAP HANA as much as possible without giving up the flexibility and extensibility of algorithms in R.

SAP BusinessObjects Predictive Analytics provides a graphical expert mode to use PAL algorithms and supports the intermixing of PAL and R algorithms within the same workflow. It is therefore SAP’s recommended predictive analytics tool for data scientists using SAP HANA.

SAP BusinessObjects Predictive Analytics is the recommended front end for all predictive technologies from SAP – including APL, PAL, and R.
“Automated” Predictive Capabilities: Using APL

The automated analytics interface within SAP BusinessObjects Predictive Analytics provides data analysts and data scientists with automated machine learning capabilities and can create predictive models without requiring data science experience (see Figure 15). It also does not require a complex predictive model as input; it simply needs to be configured and told what type of data-mining function needs to be applied to the data.

AUTOMATED PREDICTIVE MODELING CONCEPTS
If you are not in a data scientist role, you are typically analyzing data to solve specific problems related to your job. An automated machine learning system enables you to focus on the business problem you are trying to solve instead of algorithmic selection, model creation, and other predictive workflows.

With SAP BusinessObjects Predictive Analytics, this process is completely automated and therefore puts the following capabilities in the hands of almost all business users, whether they consider themselves analysts or scientists:

• Classification and regression models
• Clustering models
• Time-series analysis models
• Recommendation models

Figure 15: The Role of Business Users and Data Analysts

THE WHAT

Exploratory predictive analytics
Interact with published predictive analysis

THE HOW

Process
Create automated predictive models and simulations with easy-to-use UI
As Figure 16 illustrates, you set up the parameters for analysis so that the system can “train” on an input data set. Using sophisticated techniques for automated machine learning, SAP BusinessObjects Predictive Analytics composes its own models while creating and selectively eliminating metadata as required to create the most optimal and robust model algorithmically. You can then apply this model to a target data set to augment it with results from the predictive calculations.

The ability to do this without requiring training in mathematics or data science means that automated predictive capabilities can be put into the hands of more users. This in turn increases the value of your data platform and creates additional predictive insights with very little effort.

**Automated Predictive Analysis for Data Scientists**

Data scientists may naturally be skeptical of a fully automated predictive process. Manual modeling using a traditional predictive analysis tool can provide a skilled data scientist with more flexibility and configurability. However, these manual methods also require a significant investment in time to profile the data, create analytical models, and validate the efficiency of these models. These are all things that SAP BusinessObjects Predictive Analytics can do within a few minutes.

For most predictive scenarios, automated predictive analytics typically has comparable accuracy to hand-coded models but requires a fraction of the time. Data scientists can use automated predictive analytics to better understand their data, identify key influencers, and perform an analysis quickly. They can then decide if this level of analysis is acceptable or can continue with a manual analysis in the predictive tool of their choice – but now with a benchmarked, automated predictive model to compare with.

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**Figure 16: Simplifying the Predictive Analysis Workflow with Automation**

- Generate an analysis model
- Train the model on an input data set
- Apply the trained model to your application data set
Operationalizing Automated Predictive Models
You can export models generated from the automated analytics interface of SAP BusinessObjects Predictive Analytics as coded scoring equations in vendor-specific SQL, JavaScript, C++, Java, and numerous other formats. When an SQL-formatted scoring equation is embedded within a database, any application accessing that database can get on-the-fly in-database predictive scoring without the users being aware that a predictive calculation was done. Accessing applications can include SAP BusinessObjects BI solutions, SAP BusinessObjects Lumira, or practically any application accessing the database.

This level of automation and operationalization is a boon for developers as well, because they can directly embed the autogenerated predictive code in their applications without any significant investment in data science skills.

SAP BusinessObjects Predictive Analytics also has a server component that lets users publish their models and automate them to run without requiring the desktop tool. In this case, processing is done in the predictive server and therefore incurs the same data-transfer consequences as in the R server use case. However, this approach is database agnostic and does not rely on any calculation capabilities of the database itself.

THE AUTOMATED PREDICTIVE LIBRARY: APL
Since 2015 the APL has made these automated machine learning capabilities native to SAP HANA. Implemented as an SAP HANA–native C++ library sitting on top of the AFL, the APL features all the “automated” benefits described above but performs all calculations in-memory. This means that no data extraction is required. As part of a native library in SAP HANA, APL functions can also be called directly through SQLScript (see Figure 17).

SAP BusinessObjects Predictive Analytics speeds the creation of more accurate models by automating many of the steps typically performed by a data scientist.
Figure 17: Sample SQLScript to Call APL Functions

```sql
-- Create AFL wrappers for the APL function
-- The AFL wrapper generator needs the signature of the expected stored procedure

create column table CREATE_MODEL_SIGNATURE like PROCEDURE_SIGNATURE_T;
insert into CREATE_MODEL_SIGNATURE values (1, 'USER_APL','FUNCTION_HEADER_T', 'IN');
insert into CREATE_MODEL_SIGNATURE values (2, 'USER_APL','OPERATION_CONFIG_T', 'IN');
insert into CREATE_MODEL_SIGNATURE values (3, 'USER_APL','ADULT01_T', 'IN');
insert into CREATE_MODEL_SIGNATURE values (4, 'USER_APL','MODEL_BIN_OID_T', 'OUT');
insert into CREATE_MODEL_SIGNATURE values (5, 'USER_APL','VARIABLE_DESC_OID_T', 'OUT');

-- Call the AFL wrapper function to create APL function

call SYS.AFLLANG_WRAPPER_PROCEDURE_CREATE('APL_AREA','CREATE_MODEL','USER_APL',
'APLWRAPPER_CREATE_MODEL', CREATE_MODEL_SIGNATURE);

-- Create the input/output tables used as arguments for the APL function

create table FUNC_HEADER like FUNCTION_HEADER_T;
insert into FUNC_HEADER values ('Oid', '#42');
insert into FUNC_HEADER values ('LogLevel', '8');
insert into FUNC_HEADER values ('ModelFormat', 'bin');
create table CREATE_CONFIG like OPERATION_CONFIG_T;
insert into CREATE_CONFIG values ('APL/ModelType', 'regression/classification');
create table MODEL_BIN like MODEL_BIN_OID_T;
create table VARIABLE_DESC_OUT like VARIABLE_DESC_OID_T;

-- Execute the APL function using its AFL wrapper and the actual input/output tables

call APLWRAPPER_CREATE_MODEL(FUNC_HEADER, CREATE_CONFIG, APL_SAMPLES.ADULT01,
MODEL_BIN, VARIABLE_DESC_OUT) with overview;
```
Beyond the basic script-based approach, the APL can also be consumed from within the add-on for application function modeling in the SAP HANA studio to create AFL wrapper procedures that you can reuse (see Figure 18).

Figure 18: Using SAP HANA®, Add-On for Application Function Modeling

- Model editor
- Drag-and-drop functions
- Template for table type
- Data source selection and automatic mapping to table types

Legend
APL = automated predictive library
SAP BusinessObjects Predictive Analytics is the recommended tool to consume the APL because it can facilitate performing predictive computations directly on SAP HANA. While using the power of SAP HANA, users can perform all calculations within the database and avoid data extraction even during the initial modeling process. SAP BusinessObjects Predictive Analytics supports any combination of R, PAL, and APL algorithms in a single graphical environment when its expert interface is used (see Figure 19). Therefore, it should be the primary tool for any predictive modeling within SAP HANA.

Figure 19: Expert Mode of SAP® BusinessObjects™ Predictive Analytics
Comparing R, PAL, and APL Capabilities

Because SAP HANA supports multiple execution engines, users can choose the best predictive library for the task – and even combine usage of multiple libraries at once. It is important to realize that R, PAL, and APL are complementary technologies to provide a choice in how to plan out projects. The following sections discuss the decision criteria for choosing which library or combination of libraries would be most appropriate for a specific use case.

Note: SAP recommends deploying APL and PAL in all SAP HANA platform deployments that require predictive functionality. Customers who need additional functionality beyond what PAL provides can also deploy an R server in a sidecar configuration.

ALGORITHMIC VERSUS AUTOMATED PREDICTIVE: R AND PAL VERSUS APL

The choice between algorithmic (R and PAL) and automated (APL) predictive capabilities largely is dependent on the target users and their needs. The APL provides flexibility to automate the predictive analytics workflow without users needing knowledge of how to build complex data models from scratch. PAL or R typically requires a user to create procedures manually for each stage of the predictive modeling workflow.

Data scientists naturally tend to prefer algorithmic techniques that offer a high degree of control and precision in the modeling process. This flexibility comes at a cost: both R and PAL require users to be properly trained in data science techniques, as they must understand what each algorithm does, how it works, and how to interpret the results. Even seasoned data scientists working on less-sophisticated analyses need to invest time to go through the full predictive analytics workflow on each problem when using algorithmic techniques.

Automated analytics automates many of the predictive-modeling steps that a data scientist typically performs for common workflows like classification, regression, and association analysis, saving the user time and effort. The automated machine learning engine still performs the full predictive analytics workflow but requires very little input from the user. The result is a significantly faster analysis that has fewer configuration parameters.

In general, both data scientists and business analysts should start their analysis by using the automated predictive capabilities of SAP HANA whenever possible. Automated machine learning can address a growing number of scenarios and typically can produce valid results in seconds or minutes. This enables those who are not data scientists to answer their own questions and quickly iterate on the results in a self-service manner while giving data scientists an automated way of analyzing many problems quickly.

In some cases where a data scientist may want to create a more complex model or be in complete control of each algorithmic parameter, it is appropriate still to start with APL. That way, the data scientist can understand the data and create hypotheses before transitioning to an algorithmic method such as SAP HANA–native PAL or offboard R scripts.
**PAL AND APL VERSUS R**

Choosing SAP HANA as your predictive analytics platform gives you unrivaled capabilities to perform native in-memory analytics with any combination of algorithmic (PAL) and automated (APL) capabilities. Both of these libraries run natively in the AFL layer of SAP HANA and have direct access to the data. Calculations are performed within SAP HANA, and therefore no data is extracted, no external I/O load is created, and no external systems are required.

For cases in which a data scientist wants to use open-source, third-party, or even self-created R scripts, SAP HANA supports an external R server as a sidecar deployment. This enables practically any R script to run on data from SAP HANA, but it requires you to extract the data, process it externally, and have the results sent back to the source (see Figure 20).

Whenever possible, use technologies native to SAP HANA such as the PAL and the APL. Use an external R Server only as a last resort and for predictive calculations. Some data scientists may prefer to do data manipulation in R as well, but this would be very inefficient compared with implementing those data manipulations within a predictive tool or with a table or view from SAP HANA.

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**Figure 20: Using an External R Server for Data Sets from SAP HANA**

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Legend

- **APL** = automated predictive library
- **PAL** = predictive analysis library
Why SAP BusinessObjects Predictive Analytics with SAP HANA

SAP BusinessObjects Predictive Analytics brings a new level of predictive analytics capabilities to your organization. A wide spectrum of users from business analysts to data scientists can quickly build automated predictive models to achieve on-the-fly scoring, perform real-time predictive analysis, and embed models within BI workflows or any other business application. A single platform enables everyone to benefit from an automated predictive solution that encompasses the full predictive lifecycle, from model creation to validation and deployment, without sacrificing speed, power, or scalability.

With SAP BusinessObjects Predictive Analytics on the desktop, you gain a graphical desktop environment that supports all predictive technologies from SAP. It also ends the need to learn complicated SQLScript coding or use developer-level tools to analyze data in SAP HANA. In most scenarios, a user does not require a mathematical or data science background to start creating predictive models right away.

If you have predictive needs, start evaluating SAP BusinessObjects Predictive Analytics with SAP HANA. Take advantage of a solution that has a unique automated predictive library and model management capabilities.

Please contact your account executive for more information on how to evaluate SAP BusinessObjects Predictive Analytics, which is available for a 30-day trial at www.sap.com/trypredictive.

Review the blogs and tutorials on SAP® Community to get familiar with SAP BusinessObjects™ Predictive Analytics with SAP HANA®. In addition, visit http://help.sap.com/pa for the reference guide for the automated predictive library. For more information, visit www.sap.com/scn-predictive.