TOP TRENDS in ANALYTICS TODAY
**EIGHT KEY TRENDS in DATA ANALYTICS**

**Data analytics is** no longer the luxury of organizations with large budgets that can accommodate roving teams of analysts and data scientists. Every organization, no matter the size or industry, deserves a data analytics capability. Thanks to a convergence of technology and market forces, that’s exactly what’s happening.

Factors transforming data analytics in recent years—making it available across the board—include the cloud, the availability of open source platforms and frameworks, and the emergence of data-intensive technologies such as AI, machine learning, and the Internet of Things. This year, these trends are all converging, elevating data managers to expanded roles as advisors to and leaders of their businesses.

Trends shaping the data analytics landscape—as well as the careers of people working with data—include the following:

1. **DATA IS BECOMING STRATEGIC**
   
   Data managers are seeing their mandates extend well beyond their original roles as administrators and trouble-shooters in managing and securing day-to-day transactions. It’s now about leveraging information to make strategic, operational, and tactical decisions that result in increased revenue, improve operational efficiency, and enhance customer experience.

2. **CUSTOMER SERVICE IS BEING DRIVEN BY DATA**

   Superior customer experience, or CX, has increasingly been acknowledged as a competitive differentiator and is driving a closer alliance between marketing, customer service, and IT or data managers. That’s because CX almost wholly depends on the intelligence and insights provided by data. In addition, personalization—in which data is sourced to deliver targeted customer experiences—has emerged as a leading priority on business leaders’ agendas. The “experience” aspect doesn’t just stop at customers either—data needs to deliver highly personalized experiences to end users such as employees.

3. **MORE SELF-SERVICE**

   There is a growing emphasis on enabling end users to create their own queries to ask any question at any time of their data—without having to requisition reports from their IT departments. Vendors recognize the need for such flexibility, as well as the challenge of finding data across enterprises, and are responding with user-empowerment tools and offerings. From an organizational point of view, enterprises need to ensure that the process for acquiring and integrating this data is well-managed and well-governed, with policies and frameworks that assure the information users generate is trustworthy and reliable.

4. **THE DRIVE TOWARD DATA QUALITY INTENSIFIES**

   As data has become ever-more critical to the business, the need for decision makers to be able to trust their data has grown. Data quality is a process that begins as data sources are identified and accessed, extending to managing and storing the data. There have long been robust tools and applications...
on the market to help ensure data quality at the granular level, in terms of deduplication and cleanliness, but an effective data quality effort also needs to encompass its timeliness and the governance policies that surround it. Decision makers need to be assured that their real-time analytics and AI systems are employing the highest-quality data available.

5. DATAOPS AND DEVOPS

Not only does data need to be readily available for decision making, but a process needs to be put in place that ensures that it is moved and processed on a continuous basis, as automatically as possible. That’s why many enterprises are turning to emerging practices such as the DataOps and DevOps models of continuous integration and continuous delivery. DevOps is concerned with the process for developing and delivering application releases, while DataOps is an automated, process-oriented methodology to improve the quality and reduce the cycle time of data analytics. Lately, both of these methodologies are being applied to enable organizations to move quickly to have access to the latest algorithms and data to stay on top of their markets.

6. AI AND MACHINE LEARNING

No discussion of the power of data analytics in 2019, of course, can take place without including AI and machine learning. For starters, AI is replacing standard BI reporting as we’ve known it for decades, with real-time insights and updates on changes or developments within selected areas of the business. With AI, of course, many analytics-driven decisions can be made and executed without human intervention. At the same time, at a higher level, AI is assisting decision makers in understanding what data is telling them.

7. BOTS AND DIGITAL WORKERS

Another trend on the horizon for data analytics is the use of robotic process automation and the rise of digital workers. As more “bots” take on the day-to-day tasks of back-office work—such as managing workflows or searching files—data analytics is increasingly laying the groundwork for their intelligent performance.

8. DATA ECOSYSTEMS EXPANDING

Data environments are no longer insular systems contained within corporate walls. The ability to deliver and act on data-driven insights is increasingly amplified by connected ecosystems of partners, customers, and other constituencies. Data-driven enterprises are learning to bring together expertise and knowledge from both inside and outside their corporate walls to deliver growth and innovation.

One thing is clear: When it comes to all the possibilities data analytics offers, 2019 is a year of transformation. Data has never been more closely tied to the success of businesses, which means new opportunities for growth and advancement among the data professionals who are leading the way.

—Joe McKendrick
Today’s Analytics Challenge:
Bringing Real-Time Insight to Transactional Applications

**Like everyone else,** you have databases for transaction processing (OLTP) and data warehouses for analytics processing (OLAP). The databases support customer-facing applications. The data warehouses support internal BI/analytics teams. However, while analytics has become a competitive differentiator in every industry, it is often restricted to data warehouses and the internal BI/analytics teams using them (and other tools).

In the beginning, businesses began creating competitive differentiation by using analytics to derive valuable insight, and using it to improve operational efficiency, mitigate risk, and more. Now, the challenge is using analytics within customer-facing applications to create competitive differentiation by improving customer engagement and experience—by providing customers, rather than the business, with valuable insight (directly and/or indirectly).

This type of analytics, needed by customer-facing applications, falls somewhere between databases and data warehouses. It does not rise to the level of needing a data warehouse, but it exceeds the analytical capabilities of databases. This is the problem. For example, you may use a data warehouse for market basket analysis and the database to look up product information and store purchases, but what if you want to alert customers about “soon-to-be-sold out” products based on the products in shopping carts and recent purchases and their current inventory?

The database is great for standard CRUD operations such as viewing a product, adding it to a shopping cart, and completing a purchase—as well as for searching, filtering, and sorting products. These queries read or write most, if not all, columns in a row, and use indexes. The data warehouse is great for performing a market basket analysis to improve product recommendations. This type of query reads a small number of columns in a row, and it will not benefit from indexes—it accesses every row.

So, where do customer-facing applications send queries to sort, filter, and aggregate data in many rows without the help of indexes? These queries may need to aggregate a column for many rows in one table and use the results to look up a small number of rows in a different table. They are both column-oriented and row-oriented. For example, to alert customers of the top five “soon-to-be-sold out” products based on the quantity in active shopping carts, the quantity in recent purchases (say the last 24 hours), and quantity in current inventory. This query will use sorting, filtering, and aggregation to find the product ids for the top five “soon-to-be-sold out” products—it’s column-oriented and analytical. It may access thousands, if not millions, of shopping carts and purchases—all to get the ids of five products. It will then use them to look up five rows in the products table by product id—row-oriented and transactional.

The solution is a database capable of four things. First, storing table data in both row and columnar storage. Second, synchronizing row storage with columnar storage so all writes to row storage are replicated to columnar storage. Third, joining a table with row storage to a table with column storage. And fourth, routing queries to tables based on query type (transactional or analytical) and the underlying storage format (row or columnar) of the tables. For example, a query to get to display a shopping cart should access the row storage while a query to analyze the contents of many shopping carts should access the columnar storage.

This is the architecture of MariaDB Platform, an enterprise open source database for transactional, analytical, and hybrid transactional/analytical workloads. It uses the InnoDB storage engine for transaction processing with row storage and the MariaDB ColumnStore storage engine for analytics processing with columnar storage, synchronizes the two via streaming change-data-capture, supports cross-engine joins, and performs dynamic query routing based on configurable rules and syntax. It’s engineered to provide transactional applications with modern analytics—to deliver real-time insight to customer-facing applications.

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