

An abstract 3D rendering of a road that splits into two paths, each leading to a complex, tangled structure of overlapping road segments. The road is dark grey with white dashed lines. The background is a bright blue sky with soft white clouds. A bright light source is visible at the horizon where the two paths meet, creating a lens flare effect.

Choosing the Right In-Memory Computing Solution

A GridGain Systems In-Memory Computing White Paper

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As in-memory computing (IMC) gains momentum across a wide range of applications, from fintech and ecommerce to telecommunications and IoT, many companies are looking to IMC solutions to help them process and analyze large amounts of data in real time. However, the variety of IMC product categories and solutions can be confusing, making it difficult to determine which solution is best for a particular application. Questions abound: In-memory database or in-memory database “option”? In-memory data grid or in-memory computing platform?

This white paper reviews why IMC makes sense for today’s fast-data and big-data applications, dispels common myths about IMC, and clarifies the distinctions among IMC product categories to make the process of choosing the right IMC solution for a specific use case much easier.

Why IMC Is Right for Today’s Fast-Data and Big-Data Applications

Many data problems can be split into two groups: fast data and big data. Fast data generally applies to online transaction processing (OLTP) use cases that are characterized by high throughput and low latency. It’s important for OLTP solutions to be able to process hundreds of thousands of operations per second with latencies in milliseconds or nanoseconds. This represents operational data sets that must be consistent with reliable performance.

There’s also the big data side of the world, which is characterized by online analytical processing (OLAP) applications. In OLAP scenarios, high throughput isn’t important as they are designed for analyzing data and frequently process data offline in a batch-oriented fashion. Low latency is still critical in OLAP scenarios because we want queries to run fast.

Before focusing on IMC product categories, let’s review what IMC is and why now is the right time for memory-first data solutions.

What IMC is. In-memory computing involves storing data in RAM instead of on disk, optimally across a cluster of computers that can process the data in parallel (though some IMC product categories are more limited). Ideally, this architecture provides a way to compute and transact on data at speeds that are orders of magnitudes faster than disk-based systems, providing response times in the range of milliseconds to nanoseconds. IMC can also provide other important benefits such as scalability and data consistency, depending on the product and category.

How storage technology evolved toward today’s IMC. While in-memory technology has been around in some capacity for decades, its early forms were expensive and had limitations that made other storage technologies preferable: tape drives in the 1950s, hard drives in the 1970s, and flash drives in the 2000s. Not until the 2010s did two factors combine to make in-memory computing the technology of choice. Those two factors were RAM cost (now reduced to less than a penny per megabyte, compared to \$35 per megabyte in the mid-1990s) and the advent of 64-bit processors.

The 32-bit processors of the past had a significant limitation: they could address only four gigabytes of RAM. Once the 64-bit CPUs came in, they could address much larger spaces: multiple terabytes of RAM. This ability to put much more data in RAM gave a big push to in-memory computing.

Why IMC performs better than disk-based architectures. The reason that memory-first architectures perform faster than disk-first architectures has to do with the way the data is accessed. Disk-first architectures use disk as primary storage and use memory only for caching, while memory-first architectures use memory as primary storage and disk for backup. This is important because there are several reasons why disk access is slower than memory access.

With disk-first architectures – such as RDBMS and NoSQL databases and file systems such as Hadoop’s HDFS – there is a multi-step process that must occur in order to access a byte of data. The request starts with an API call, then it goes to the operating system and through the I/O controller before finally accessing the disk to load the data. At this point, even if only one byte of data is needed, more must be loaded, because disk is block-addressable and not byte-addressable storage. This is significant because an entire block (whichever block size has been configured in the OS, typical configurations for a data store volume could be four kilobytes or eight kilobytes) must be loaded. Each step introduces latency, from several milliseconds to hundreds of milliseconds.

With a memory-first architecture, the process is much simpler and faster. Accessing a byte of data requires only the referencing of a pointer (which specifies the location of that byte in memory). No loading is needed. The system can read kilobytes of data immediately without loading it or processing it. There is direct access to any portion of the memory, and that access is measured in nanoseconds not in milliseconds like the disk access described above.

For applications requiring truly fast performance – nanoseconds or microseconds instead of milliseconds – memory-first architectures are the clear choice.

Dispelling Myths About IMC

Of course, while performance is a key factor in deciding on a data architecture, there are other factors as well – and sometimes myths about in-memory computing come into play when those factors are considered. Let’s look at four common myths about in-memory computing and why they are not based in fact.

Myth #1: Too expensive. In the mid-1990s, loading a terabyte of data in-memory cost millions of dollars. However, the cost of RAM has been decreasing about 30% per year since then, so it is now much more affordable. Recently, Sberbank found that they could generate one billion transactions per second with the GridGain in-memory computing platform using only 10 Dell blades with a combined memory of one terabyte – at a cost of about \$25,000. So, memory is now very affordable and it is being used as a main system of records in many use cases.

Myth #2: Not durable. While it is true that DRAM will not survive a server crash, most of today's systems are designed to solve that problem through fault-tolerance strategies such as automatic data replication and integration with persistent data stores. For example, the Apache® Ignite™ in-memory computing platform (and its commercial version, GridGain), will automatically persist and load data from any RDBMS or NoSQL database on disk.

Also, there are some promising developments on the horizon in the area of non-volatile memory, such as Intel's 3D XPoint: byte-addressable, flash-based memory that will plug into PCI slots and act almost like a regular DRAM. Because of its non-volatile nature, this type of memory will be able to survive several restarts. While it will not be as fast as DRAM, it will be faster than today's flash.

Myth #3: Flash is fast enough. While today's flash-based memory is definitely faster than spinning disks, it is still generally a block-level device and still must go through OS I/O, the I/O controller, marshaling, and buffering. For applications that need to access individual bytes of data, DRAM is still about a hundred times – or, in some cases, a thousand times – faster than flash-based access. If a ten-millisecond latency is good enough for a particular application, then flash-based memory is fine. However, if the application requires latency of one millisecond or half a millisecond, there is a benefit to paying a little more, and doing most of the processing through DRAM.

Myth #4. Memory is only for caching. While caching is an important aspect of memory, many systems use caching only as a small component of their architectures. Today, many types of processing happen directly in memory: machine learning, streaming, queries, transactions, and so on.

As awareness grows that these common misperceptions are not based in fact, interest in in-memory computing products is growing substantially.

IMC Product Categories

There are many different terminologies for types of in-memory computing. It's important to understand where they came from, their capabilities, and how they can be unified into a complete in-memory computing platform.

In-memory computing products generally fit into the following four categories:

- In-memory database options
- In-memory databases
- In-memory data grids
- In-memory computing platforms

The first of these categories, in-memory database options, is very limited in scope, and not scalable to fit with the distributed architectures of the future. The remaining three categories involve distributed-memory architectures, so they are better suited to the fast-data OLTP applications that are becoming more prevalent today. These categories also include some products that can capably address big-data OLAP issues by integrating with products such as Apache Hadoop®, Apache Spark®, and others.

The table “IMC Product Categories” on page 8 provides an overview of the advantages and disadvantages of the four product categories, which are described in more detail below.

In-memory database options. This category describes configuration options for existing databases that improve their performance by increasing the amount of data accessed in memory rather than on disk. Such options are available with Oracle® Database 12c, Microsoft SQL Server®, and many other RDBMS products.

The main advantage of using one of these products is that it works with a company’s existing database, with no need to migrate data, change APIs, or change any code. Generally, all that is needed is a configuration change and the addition of more memory. Database performance will improve immediately.

However, this is a very limited solution, because it involves using the memory on only one server. There is no scalability, unlike other in-memory solutions that scale out in a distributed fashion and allow access to the total memory available on all of the servers in a cluster.

In-memory databases. Products in this category, such as MemSQL® and VoltDB®, behave just like databases that keep all of their data in memory (distributed across multiple servers) for fast access. While they have some persistence capabilities, they are basically memory-first architectures. They are called databases because the way to interact with them is to use SQL with standard database drivers. Users can create schemas, issue queries, and use DDL or DML statements such as inserts, updates, deletes, and so on.

These systems have very high throughput because the data is accessed in memory. They also exhibit good scalability because they distribute their data across multiple servers. However, they also have some significant disadvantages.

With an in-memory database, SQL is the only way to talk to the system, and SQL does not fit into all distributed-use cases. Sometimes it is important to be able to perform computations directly on the server where the data resides in order to optimize performance.

Another major disadvantage is that a company’s existing database must be set aside to install the new in-memory database and that becomes the system of record. Most companies have a lot of infrastructure code around their existing databases, so migrating to a new database requires a massive effort. For this reason, in-memory databases are most feasible for companies that are either selecting a first database or needing to completely replace their existing database. Companies that are instead seeking to accelerate their existing databases are generally better off with products such as in-memory data grids or in-memory computing platforms.

It’s worthy to note that Apache Ignite / GridGain, a full-featured in-memory computing platform also fully supports ANSI SQL-99 including DML and DDL.

In-memory data grids. This group includes products such as Hazelcast®, GigaSpaces®, Oracle Coherence®, Apache Geode® / Pivotal GemFire®, and Infinispan®. In some cases, like that of Apache

Ignite / GridGain, this category is a subset of the larger category of in-memory computing platforms (IMCP). IMCPs include in-memory data grids among their other in-memory components.

Like in-memory databases, products in this category also distribute data across multiple servers. However, the methods of accessing the distributed data are different: data grids typically provide data access through a key-value API and sometimes a custom query API, with limited or nonexistent SQL support. In addition, data grids generally provide the ability to collocate custom computations for specific data.

The ability to collocate computations with data provides some significant performance advantages. A typical application might involve, for example, accessing many accounts and closing their balances or performing other computations on them. By performing such actions directly on the server where the data resides, data grids can provide very high throughput and very low latencies. In a distributed system, the more the system can collocate computations with their associated data, the faster and more scalable the system will be because this eliminates network bottlenecks.

The down side of this architecture is that it is necessary to develop code to integrate the logic with the data grid. However, the benefit is extremely fast performance, with hundreds of thousands of operations per second throughput.

Another major benefit that data grids offer is compatibility with existing databases. Unlike in-memory databases, which must become the system of record, in-memory data grids can be inserted on top of an existing database and integrate with it. For example, if a company already has a database installed – such as Oracle, Microsoft SQL Server, or Apache Cassandra[®] – the data grid provides an API that fits between the database and the application, providing large amounts of caching and data distribution.

Because not all data-grid products are created equal, it's important to look for the following features:

- **JCache implementation.** Java garbage-collection pauses can add significant time delays in situations where hundreds of gigabytes of data are stored across clusters containing tens of nodes. Data grids that implement the JCache standard, JSR 107, can avoid the garbage-collection process by storing and managing data outside of Java in off-heap memory. For better performance, choose a data grid that implements the JCache JSR 107 standard, such as the data grid component of the in-memory computing platform Apache Ignite / GridGain, Oracle Coherence, and the enterprise version of Hazelcast.
- **Transactional consistency.** Managing transactions across a distributed grid is more challenging than managing them in a non-distributed architecture. In choosing a data grid, consider what level of control it offers over transactional consistency. This is particularly important for financial applications. Eventual consistency is not as strong a guarantee as fully ACID-compliant transactions. Look for a data grid with strong transactional semantics, such as the data grid component of the in-memory computing platform Apache Ignite / GridGain.
- **Full SQL support with in-memory indexing.** Some data grids offer much more substantial query support than just providing key-value access to the data. Some products provide limited SQL or custom APIs for querying data, and one – the data grid component of the in-memory computing platform

Apache Ignite / GridGain – supports full ANSI SQL-99, including DDL and DML. This support lets companies leverage their existing SQL code through JDBC and ODBC APIs. Also, Ignite / GridGain stores indexes in-memory, which is important because indexes substantially improve performance of queries and joins.

Another important point to remember in choosing a data grid is that some data-grid products include other important in-memory components in addition to the data grid, placing them in the more encompassing category of in-memory computing platforms.

In-memory computing platforms. Anyone seriously evaluating in-memory computing should focus their attention on the ultimate form of this technology: in-memory computing platforms. With in-memory technology in particular, the platform approach holds greater value than the point solution approach. Major data centers don't want to install and manage myriad point solutions. They want a single platform that supports all use cases. They want a strategic approach to in-memory computing, and that solution is an in-memory computing platform.

The category of in-memory computing platform includes data-grid products such as Apache Ignite, GridGain (the enterprise version of Apache Ignite), and Terracotta®. While data grids are essentially distributed caches with some basic transactional and compute capabilities, featuring key-value data access, in-memory computing platforms are more comprehensive products with much more extensive capabilities.

In the case of Apache Ignite and GridGain, the extra features putting these products in the category of in-memory computing platform include the following:

- **Multiple, well-integrated in-memory components:** Apache Ignite and GridGain include not only a data grid and compute grid, but also a SQL grid, service grid, a file system, and a streaming component – all in-memory, all well integrated with each other. For example, computations can be sent to the nodes where the data is, and a service can be attached to the data as well. With streaming, the data is streamed to the node where it will be eventually stored and processed. All of these components know how to work with each other automatically, without any additional effort from the user. GridGain Enterprise Edition integrates additional features with these components, including security management and monitoring, administration, data-center replication, and network-segmentation resolution.
- **Multiple ways to talk to data:** Data access is another area that distinguishes in-memory computing platforms from in-memory databases, which support data access only through SQL, and in-memory data grids, which provide data access primarily through a key-value API. In-memory computing platforms provide multiple ways to talk to the data. For example, Apache Ignite and GridGain provide both ANSI SQL-99 and key-value access, plus a wide range of other access mechanisms via a Unified API, including C++, .NET, Java, PHP, and ODBC and JDBC drivers, which allow integration with data visualization tools, such as Tableau®. Plus, the Ignite file system integrates with the Hadoop file system, HDFS, for persistence to memory instead of disk, and there is a pluggable, in-memory MapReduce component that integrates with Spark for additional big-data cluster-computing power.

Thanks to features such as these, in-memory computing platforms provide performance and scalability improvements and flexibility not available with a product that is simply a data grid. For example, with Apache Ignite and GridGain, data might be stored using the key-value API and accessed using the SQL API, or stored in Java and accessed in .NET – all with the streamlined performance of multiple integrated in-memory components.

Table of IMC Product Categories

Category	Description	Advantages	Disadvantages	Best Suited For
In-memory database option Examples: Oracle Database 12c, Microsoft SQL server, and others	Configuration option for an existing database that boosts performance by caching data in memory	Works with existing database	Not scalable; Limited to memory on one server; Not a distributed architecture	Companies that want to improve performance of an existing database and don't anticipate scaling to work with large volumes of data at real-time speeds
In-memory database Examples: MemSQL, VoltDB, SAP HANA®	RDBMS that stores data in memory instead of on disk	High throughput; High scalability; Disk persistence (disk is a copy of memory); Full SQL support; Applications can remain unchanged	Requires complete replacement of current database; Can interact only using SQL; no distributed parallel processing	Companies in the process of selecting a first database that require speed for fast-data (OLTP) situations
In-memory data grid Examples: Hazelcast, GigaSpaces, Oracle Coherence, Apache Geode / Pivotal GemFire, Infinispan	Distributed in-memory key-value store	High throughput; Low latency; Very high scalability; no need to replace existing databases	Requires code to integrate logic with data grid; key-value stores may not include complete SQL support	Companies requiring very high speed and scalability for fast-data (OLTP) applications and the flexibility to work with either new or existing databases
In-memory computing platform Examples: Apache Ignite / GridGain In-Memory Computing Platforms, Terracotta	Includes an in-memory data grid plus several fully integrated in-memory components and features	Same as in-memory data grid, plus: Multiple ways to access data; Integrated additional components	Requires code to integrate logic with data as it is stored in the in-memory computing platform	Companies that require a full-featured platform with support for OLTP plus SQL-based OLAP and HTAP applications

Sberbank Case Study: In-Memory Computing in Action

One of the most noteworthy GridGain Systems financial services customers is Sberbank, the largest bank in Russia and the third largest in Europe. Sberbank was faced with a similar problem to the one currently facing companies engaged in building a modern in-memory computing architecture. The bank was switching from a more traditional, brick-and-mortar setup – one in which people would come into their offices and manually process a limited number of financial transactions each day, during a limited time period – to a new world with online and mobile customers transacting with them 24/7.

The company forecasted future throughput requirements and determined that it needed to move to a next generation data processing platform to handle the expected transaction volume. Sberbank analyzed more than ten potential solutions from vendors in the in-memory computing space and found that the GridGain in-memory computing platform was the most comprehensive solution. The bank concluded that GridGain would provide the next-generation platform with a significant improvement in performance and scalability.

The GridGain in-memory computing platform provided several other important capabilities that Sberbank's next-generation platform would require such as machine-learning and analytics, flexible pricing, artificial intelligence, ease of deployment, hardware independence of cluster components, and a rigorous level of transactional consistency. Of particular importance was the ability to conduct integrity checking and rollback on financial transactions. Sberbank could not find that level of consistency with other in-memory computing solutions.

In a [January 2016 article in RBC](#), Herman Gref, the CEO of Sberbank, said that the bank selected the GridGain Systems technology to build “a platform that will enable the bank to introduce new products within hours, not weeks.” He went on to state that the GridGain in-memory computing platform enables Sberbank to provide “unlimited performance and very high reliability” while being “much cheaper” than the technology used previously. Sberbank is using GridGain's in-memory computing platform to implement capabilities that could not be provided by the other vendors evaluated – a group that included Oracle®, IBM® and others.

GridGain Systems: A Leader in In-Memory Computing

With companies grappling with the challenges resulting from increasing fast-data and big-data workloads, demand for the GridGain in-memory computing platform is growing dramatically. This comprehensive platform contains a complete feature set that surpasses the capabilities of in-memory database point solutions, making it well suited for OLAP, OLTP and HTAP use cases.

As a complete in-memory computing platform, GridGain helps users consolidate onto a single high performance and highly scalable big-data solution for transactions and analytics, resulting in lowered TCO. Advanced SQL functionality and API-based support for common programming languages enable

rapid deployment. This, coupled with the rapidly decreasing cost of memory, boosts ROI for in-memory computing initiatives, enabling businesses to build less expensive systems that perform thousands of times better.

Clients enjoy the following:

- **A unified high-performance architecture.** The GridGain in-memory computing platform consists of multiple grids connected by a clustered in-memory file system. The In-memory Data Grid, Compute Grid, SQL Grid, Streaming Grid and Service Grid are interconnected. Computations occur as close as possible to the data used in the computation. Additional features such as high throughput, low latency, load balancing, caching, in-memory indexing, streaming, Hadoop acceleration and other performance improvements are crucial to success in real-time modeling, processing, and analytics.
- **Scalability.** The GridGain in-memory computing platform excels in terms of scalability, allowing companies to add cluster nodes and memory in real-time with automatic data rebalancing. As a hardware-agnostic solution, clients can choose their preferred hardware for scaling up.
- **Full SQL support.** GridGain is ANSI SQL-99 compliant and the In-Memory SQL Grid supports DML so users can leverage their existing SQL code using the GridGain JDBC and ODBC APIs. For users with existing code bases which are not based on SQL, they can leverage their existing code through supported APIs for Java, .NET, C++, and more.
- **High availability.** The GridGain in-memory computing platform provides essential high availability features such as data-center replication, automatic failover, fault tolerance, and quick recovery on an enterprise-level scale.
- **Transaction processing.** The GridGain in-memory computing platform supports ACID-compliant transactions in a number of user-configurable modes.
- **Security features.** The GridGain in-memory computing platform supports authentication, authorization, multiple encryption levels, tracing, and auditing.
- **Open Source framework.** GridGain is based on Apache® Ignite™, a popular open source project with many contributors that has been tested globally. GridGain Systems was the original creator of the code contributed to the Apache Software Foundation that became Apache Ignite and fully supports the technology behind Apache Ignite. The [GridGain Enterprise Edition](#) extends the features in Apache Ignite to provide enterprise-level capabilities and services, such as additional security, data center replication, auditing mechanisms, a GUI for management and monitoring, network segmentation, and a recoverable local store.
- **Production Support.** [GridGain Systems Support](#), available to [GridGain Professional Edition](#) and [GridGain Enterprise Edition](#) users, includes rolling updates, faster availability of all releases and patches, and 24/7 enterprise-level support.

Making the Choice

Companies facing the need to process massive amounts of data in real time can now benefit from a robust and affordable range of in-memory computing solutions. From the simple and limited performance boost of in-memory database options to the more scalable and high-performing distributed architectures of in-memory databases (which replace existing databases), in-memory data grids (which integrate with existing databases), and in-memory computing platforms (which include data

grids and much more), there is an appropriate product category for every use case. Understanding the distinctions among these categories is the key to making the best choice for any big-data or fast-data application requiring in-memory OLTP, OLAP or HTAP strategies.

Contact GridGain Systems

To learn more about how GridGain In-Memory Data Fabric can help your business, please email our sales team at sales@gridgain.com, call us at +1 (650) 241-2281 (US) or +44 (0) 7775 835 770 (Europe), or complete our [contact form](#) to have us contact you.

About GridGain Systems

GridGain Systems is revolutionizing real-time data access and processing by offering enterprise-grade in-memory computing solutions built on Apache Ignite. GridGain solutions are used by global enterprises in financial, software, eCommerce, retail, online business services, healthcare, telecom and other major sectors. GridGain solutions connect data stores (SQL, NoSQL, and Apache™ Hadoop®) with cloud-scale applications and enable massive data throughput and ultra-low latencies across a scalable, distributed cluster of commodity servers. GridGain is the most comprehensive, enterprise-grade in-memory computing platform for high volume ACID transactions, real-time analytics, and hybrid transactional/analytical processing. For more information, visit gridgain.com.

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