EMC ISILON NAS: PERFORMANCE AT SCALE FOR ELECTRONIC DESIGN AUTOMATION

ABSTRACT
This paper describes how EMC Isilon network-attached storage delivers performance, scalability, and efficiency to optimize data storage for electronic design automation.

June 2014
# TABLE OF CONTENTS

## ABSTRACT 1

## EXECUTIVE SUMMARY 4

## EDA WORKFLOWS, INFRASTRUCTURE, AND WORKLOADS 4

- **Workflow** 4
- **Infrastructure** 6
- **Workload** 7
- **Data Growth** 7

**Traditional Storage Architectures for EDA** 8

**Problems in EDA Data Storage** 8

**Overview of Isilon Scale-Out NAS** 9

**Performance at Scale** 10

**Storage Efficiency and Utilization** 11

**Data Management** 13

## CONCLUSION 14
EXECUTIVE SUMMARY

The EMC® Isilon® platform combines a scalable architecture with distributed software to optimize data storage for electronic design automation (EDA). Isilon scale-out network-attached storage is a fully distributed, symmetrical system that consists of storage nodes arranged in a cluster. The Isilon OneFS® operating system unites the memory, I/O, CPUs, and disks of the nodes to present a global namespace as a single file system.

Each node adds capacity, performance, and resiliency to the cluster, and each node can process requests from EDA clients while taking advantage of the entire cluster’s performance. The Isilon architecture contains no single master for the data, no concept of a controller head, and no RAID groups. The result is a highly efficient, scalable architecture that overcomes problems in EDA data storage.

One problem with traditional EDA storage architectures is that they create performance bottlenecks that worsen at scale. The controller is the primary bottleneck: Attaching too much capacity to the controller can saturate it. Bottlenecks levied by the storage system reduce wall-clock performance for concurrent jobs, which can lengthen the time it takes to bring a chip to market.

The distributed Isilon architecture eliminates the single-head CPU saturation point of a controller. A large number concurrent jobs, often resulting in large amounts of metadata operations, can run without saturating the storage system, shortening the time to market.

Traditional storage systems also use disk space inefficiently. The uneven utilization of capacity across islands of storage requires efforts to manually rebalance volumes across aggregates and to manually migrate data to an even level—work that increases operating expenses. The inefficient utilization also increases capital expenditures because extra capacity must be set aside as storage overhead.

In contrast, OneFS evenly distributes data among a cluster’s nodes with layout algorithms that maximize storage efficiency. The Isilon system continuously reallocates data to conserve space and eliminate much of the capacity overhead that traditional EDA storage systems require. Isilon’s efficient utilization of disk space reduces capital expenditures and operational expenses.

Traditional storage systems have multiple points of management. Each filer must be individually managed, and the management overhead increases the total cost of ownership and OpEx. The lack of centralized management also puts EDA companies at a strategic disadvantage because it undermines their ability to expand storage to adapt to ever-expanding data sets and fluctuating business needs, which can hamper efforts to reduce time to market. With its single volume, the Isilon architecture delivers a high return on investment by centralizing data management.

By scaling multidimensionally to handle the growth of EDA data, an Isilon cluster lets you adapt to fluid storage requirements, nondisruptively add capacity and performance in cost-effective increments, and improve wall-clock run times for concurrent jobs. This paper describes how an EMC Isilon cluster delivers performance, scalability, and efficiency to optimize data storage for electronic design automation.

EDA WORKFLOWS, INFRASTRUCTURE, AND WORKLOADS

The workflows, workloads, and infrastructure for chip design—combined with exponential data growth and the time-to-market sensitivity of the industry—constitute a clear requirement to optimize the system that stores EDA data. Summarizing the workflow, infrastructure, and workload typical of EDA sets the stage for identifying problems that undermine the performance, efficiency, and elasticity of most EDA storage systems.

Workflow

Most EDA workflows include the following phases:

- Frontend design phases (logical design)
  - Design Specification
  - Functional Verification
  - Synthesis
  - Logic Verification
- Backend design phases (physical design)
  - Place and Route
  - Static Timing Analysis
  - Physical Verification
  - Tape-out

These phases interact to form a digital design flow for EDA:

During the frontend design phases, engineers architect a chip design by compiling source files into a chip model. Engineers then verify the chip design by scheduling and running jobs in a large compute grid. A scheduler distributes the build and simulation jobs to the available slots on the compute resources. The workflow is iterative as the engineers refine the design, and it is the frontend design phase that generates the majority of the simulation jobs. Efficiency in creating, scheduling, and executing build and simulation jobs can reduce the time it takes to bring a chip to market.

The frontend phase generates an input/output (I/O)-intensive workload when a large number of jobs run in parallel: EDA applications read and compile millions of small source files to build and simulate a chip design.

A storage system manages the various design projects and files so that different users, scripts, and applications can access the data. During the frontend verification stages, the data access pattern tends to be random, with a large number of small files. The frontend workload requires high levels of concurrency because of the large number of jobs that need to run in parallel, generating a random I/O pattern.

During the backend design and verification phases, the data access pattern becomes more sequential. The backend workload tends to have a smaller number of jobs with a sequential I/O pattern that run for a longer period of time.

The output of all the jobs involved in a chip’s design phases can produce terabytes of data. Even though the output is often considered scratch space, the data still requires the highest tier of storage for performance.

Within the storage system, EDA workflows tend to store a large number of files in a single directory, typically per design phase, amid a deep directory structure on a large storage system. The performance-sensitive project directories, including those for both scratch and non-scratch directories, dominate the file system.

The directories contain source code trees, frontend RTL files that define logic in Hardware Description Language (HDL), binary compiled files after synthesis against foundry libraries, and the output of functional verifications and other simulations. Typically, the
frontend RTL project directories that contain source code are minimal in size, while the project directories used for simulations dominate the overall capacity utilization of the storage system.

For example, an EDA company might have the following type of hierarchy for chip design under a top-level directory named projects:

<table>
<thead>
<tr>
<th>FUNCTION</th>
<th>EXAMPLE DIRECTORY</th>
</tr>
</thead>
<tbody>
<tr>
<td>Frontend Design Data</td>
<td>/projects/chip1_FE or /projects/chip1_RTL</td>
</tr>
<tr>
<td>Scratch Space for Simulations</td>
<td>/projects/chip1_scratch, /projects/chip1_verif</td>
</tr>
<tr>
<td>Physical Layout</td>
<td>/projects/chip1_layout, /projects/chip1_pnr</td>
</tr>
<tr>
<td>Physical Verification</td>
<td>/projects/chip1_physical_verif</td>
</tr>
</tbody>
</table>

Table 1. Typical Project Directory Layout For Chip Design Data

With an inadequate storage system, these kinds of directory structures can have implications for latency. If EDA engineers are running 500 jobs against the /project/chip1_verif directory while another user tries to interact with the directory, the user’s response time can be noticeably delayed.

The workflow for the projects requires backing up data and taking snapshots of the directories nightly, and sometimes retaining up to over a month worth of snapshots of project directories. The storage system may also include home directories for users and the safe but cost-effective archiving of design blocks for reuse in future projects.

After the frontend phase produces a design that works logically, the physical design phase converts the logical design into an image, verifies it, and prepares it for delivery to a foundry, which manufactures the chip. During the physical design and verification stage, files increase in size and the data access pattern becomes more sequential. The result is a single top-level GDS-II file that can be tens of gigabytes.

The GDS file then undergoes a process of top-level verification to ensure that the design is clean. The last phase uses software to perform multiple iterations of a design rule check (DRC) and a layout versus schematic test (LVS). The DRC makes sure that the chip's layout conforms to the rules required for faultless fabrication, and the LVS guarantees that the layout represents the circuit that you want to fabricate. For nanometer technologies, a design for manufacturing (DFM) solution also helps increase the yield.

If an EDA company lacks a storage system that can handle the multiple iterations of DRC, LVS, and DFM checks on a large file while on a tight deadline, the company might not have enough time to perform due diligence on the layout, with the result being costly delays during fabrication or a decrease in the yield.

Given the large number of files and directories as well as the growing size of design files over time and the verification tests that they require, the storage system must make file management easy while providing seamless, high-performance access.

**Infrastructure**

A typical EDA infrastructure combines a large compute farm with storage servers, application servers, license servers, backup systems, infrastructure servers, and other components, all connected with an Ethernet network. EDA uses a large compute grid, or build farm, for high-performance computing, and the grid often consists of more than a 1,000 cores.

Compute farms are increasing in density. In the past, the compute farms were noncommodity UNIX servers with a limited amount of compute cycles, such as 4 CPUs in 6U chassis—a system that was neither dense nor fast. In contrast, today the compute grid comprises hundreds of commodity Linux servers.

Storage servers are also seeing exponential growth. In the past, storage was not a bottleneck but now with denser commodity compute hardware and more cores—as many as 10,000 to 25,000 at large EDA shops—the compute grid is becoming dense and extensive, leaving storage as the bottleneck. For many EDA companies, the storage system still has a legacy dual controller head architecture, such as a traditional scale-up NAS system.
**Workload**

The clients and applications in the compute grid determine the characteristics of the storage system’s workload:

Most EDA workloads access the storage system over Network File System (NFS). Across the EDA industry, I/O profiles vary by tool and design stage, from small random workloads to large sequential workloads. In general, the workload tends to be a mix of random and sequential input-output operations.

The frontend workflow requires high levels of concurrency rather than throughput. As hundreds of functional and logic verification jobs are run concurrently against a deep and wide directory structure, the workload induces tremendous amounts of metadata overhead, generating high CPU usage on the storage system. The data access pattern is random. In many cases, most file system operations retrieve attributes, perform lookups, or access other metadata. It is the metadata intensive nature of the workload that tends to saturate the storage subsystem, making the controller the bottleneck in a scale-up architecture.

In contrast to the frontend workflow, the backend design flow consists of running a fewer number of jobs, but it requires an increase in the amount of throughput because of the increase in file sizes. The data access pattern for physical verification becomes more sequential compared to functional verification, and the workload now exhibits increased read-write operations and a sequential data access pattern.

**Data Growth**

Like all unstructured data, EDA data is growing exponentially. With each design phase of every EDA project, the data that needs to be stored expands.

![File Based Data Growth](image)

*Figure 2. Growth of unstructured data across vertical markets.*

In EDA, data growth is taking place as the technology-node transitions to smaller geometries: EDA companies are looking to 10nm and below. As designs get more complex and dense, an EDA project that used to take only hundreds of GBs can now require as much as tens of TBs of stored data as the design is prepared for tape-out to the foundry, even though the final result that is delivered to the foundry is only tens of GBs of GDS-II files.

Amid such exponential data growth and the changing face of electronic design automation, combining the characteristics of an EDA workload with its typical infrastructure makes storage the performance bottleneck.
Traditional Storage Architectures for EDA
For years, a typical EDA infrastructure has relied on the same architecture for its storage system: A scale-up storage system characterized by a single-server operating system and a controller head with shelves of disks. The architecture creates islands of storage with many disk shelves and many separate controller heads.

Traditional architectures for EDA storage also rely on a failover relationship between two controllers that form a high-availability (HA) pair in which each controller provides high availability for the other. When failure occurs, the clients, connections, and storage operations that were handled by a single storage controller are moved to a neighboring storage controller.

Problems in EDA Data Storage
Using a traditional scale-up storage architecture for EDA leads to an array of problems:

PERFORMANCE BOTTLENECKS. Traditional EDA storage architectures create performance bottlenecks—bottlenecks that get worse at scale. A traditional scale-up architecture requires manual data migrations among storage components to balance performance for EDA applications. The controller is the main bottleneck: Although it is typical in EDA to limit the amount of capacity per controller head to make sure that performance requirements are met, attaching too much capacity to the controller can saturate it—a situation made worse by the fact that adding capacity does not scale performance. Performance bottlenecks levied by the storage system can reduce wall-clock performance for concurrent jobs, which can affect the time it takes to bring a chip to market and, ultimately, revenue.

INEFFICIENT UTILIZATION OF DISK SPACE. Capacity is unevenly utilized across islands of storage: Some volumes are underutilized while others are oversubscribed. The result is many volumes, all with pockets of free space. The uneven utilization forces you to manually rebalance volumes across aggregates and to manually migrate data to an even level. The burden of managing data across volumes and migrating data to distribute it evenly not only undermines performance but also increases operational expenses (OpEx). Despite the rebalance efforts, the typical raw to usable to used ratios of storage capacity for each volume is as follows:

<table>
<thead>
<tr>
<th>RAW CAPACITY</th>
<th>USABLE CAPACITY</th>
<th>USED CAPACITY</th>
</tr>
</thead>
<tbody>
<tr>
<td>100 percent</td>
<td>65 percent of raw capacity</td>
<td>50 percent of raw capacity</td>
</tr>
</tbody>
</table>

Table 2. Typical Capacity Utilization Ratios in a Scale-up Storage System

MULTIPLE POINTS OF MANAGEMENT. With no central point of management, each filer must be individually managed. The management overhead increases the total cost of ownership (TCO) along with OpEx. The multiple points of manual management also put EDA companies at a strategic disadvantage because the lack of centralized management undermines a business unit’s ability to expand storage to adapt to demand, which can in turn hamper efforts to reduce time to market. With file server sprawl, the cost of managing fast-growing data can also exceed the IT budget by resulting in costly data migrations to eliminate hot spots. Similarly, backup and replication becomes increasingly complex and costly.

STORAGE INELASTICITY. Expanding data sets coupled with dynamic business models can make storage requirements difficult to forecast. Up-front purchasing decisions, based on business and data forecasts, can misestimate storage capacity. Forecasting capacity in advance of actual known needs undermines adaptability to changing business needs. The lack of agility can affect time to market and profit margins. Such a model of early storage provisioning can also lead to vendor lock-in and a loss of negotiation leverage that increases costs.

Isilon Solutions for EDA Workloads
Isilon overcomes the problems that undermine traditional NAS systems by combining the three traditional layers of storage architecture—file system, volume manager, and data protection—into a scale-out NAS cluster with a distributed file system. Such a scale-out architecture increases performance for concurrent jobs, improves disk utilization and storage efficiency to lower capital expenditures (CapEx), centralizes management to reduce operating expenses, and delivers strategic advantages to adapt to changing storage requirements and to improve time to market. After presenting an overview of Isilon NAS, this section details how the Isilon solution overcomes problems in EDA data storage.
Overview of Isilon Scale-Out NAS
The EMC Isilon scale-out platform combines modular hardware with unified software to provide the storage foundation for electronic design automation. Isilon NAS is a fully distributed, symmetrical system that consists of nodes of modular hardware arranged in a cluster. The distributed Isilon OneFS operating system combines the memory, I/O, CPUs, and disks of the nodes into a cohesive storage unit to present a global namespace as a single file system.

The nodes work together as peers in a shared-nothing hardware architecture with no single point of failure. Every node adds capacity, performance, and resiliency to the cluster, and each node can process requests from EDA clients simultaneously. A compute client can connect to any node with standard protocols like NFS and SMB while taking advantage of the entire cluster's resources. The architecture contains no single master for the data or metadata, no concept of centralized metadata server, and no concept of a controller head to cause bottlenecks.

As nodes are added, the file system expands dynamically and redistributes data, eliminating the work of partitioning disks and creating volumes. The result is a highly efficient and resilient storage architecture that brings all the advantages of an enterprise scale-out NAS system to storing data for EDA.

An Isilon cluster optimizes data protection. OneFS more efficiently and reliably protects data than traditional scale-up storage systems. OneFS stripes data across the cluster and protects the data with highly efficient forward error correction codes.

An Isilon cluster also includes enterprise features to back up your data and to provide high availability. OneFS can perform NDMP (Network Data Management Protocol) backups, synchronization, geo-replication, snapshots, file system journal, virtual hot spare, IntegrityScan, dynamic sector repair, and accelerated drive rebuilds. For complete information about the data availability features of OneFS, see the white paper titled Isilon: Data Availability and Protection.

The comprehensive enterprise features of OneFS ease data management. OneFS includes storage pools, deduplication, automated tiering, quotas, high-performing SSDs, capacity-optimized HDDs, and monitoring with InsightIQ. SmartQuotas, for instance, can restrict project spaces, and SmartConnect can balance client connections evenly among nodes for high performance. SmartPools provides tiered storage so that you can place current data in a high-performance storage pool while placing older data in a lower, more cost-effective pool after the project has been tape-out.

As a rack-mountable appliance, a node includes the following components in a 2U or 4U rack-mountable chassis: memory, CPUs, RAM, NVRAM, Ethernet cards, InfiniBand adapters, disk controllers, and storage media. Each node runs the Isilon OneFS operating system, the distributed file system software that unites the nodes into a cluster. An Isilon cluster comprises three or more nodes and can scale up to 144. A cluster's storage capacity ranges from a minimum of 18 TB to a maximum of 20 PB.

When you add a node to a cluster, you increase the cluster's aggregate disk, cache, CPU, RAM, and network capacity. OneFS groups RAM into a single globally coherent cache so that a data request on a node benefits from data that is cached anywhere. NVRAM is grouped to write data with high throughput and to protect write operations from power failures. As the cluster expands, spindles and CPU combine to increase throughput, capacity, and I/O operations per second (IOPS). OneFS supports NFS, SMB, FTP, HDFS, and HTTP; there is no need for client-side drivers, hence seamlessly integrating into an existing EDA infrastructure.

EMC Isilon makes several types of nodes, all of which can be added to a cluster to balance capacity and performance with throughput or IOPS.

<table>
<thead>
<tr>
<th>MODEL</th>
<th>USE CASE</th>
</tr>
</thead>
<tbody>
<tr>
<td>S-Series</td>
<td>IOPS-intensive applications such as EDA</td>
</tr>
<tr>
<td>X-Series</td>
<td>High-concurrency and throughput-driven workflows</td>
</tr>
<tr>
<td>NL-Series</td>
<td>Near-primary accessibility, with near-tape value</td>
</tr>
</tbody>
</table>

Table 3. EMC Isilon node types.

The EMC Isilon Performance Accelerator extension node provides independent scaling for high performance by adding processing power, memory, bandwidth, and parallel read/write access.

As the next sections of this paper demonstrate, the architecture of Isilon scale-out NAS is ideally suited to reduce bottlenecks, optimize concurrency, store EDA data efficiently and cost effectively, reduce OpEx through centralized management, and shorten time to market with strategic advantages.
Performance at Scale

**EFFICIENT METADATA ACCESS:** An EDA workload can be classified into three primary operations: reads, writes, and metadata access. The workload can be characterized as having a typical mix of 65 percent metadata, 20 percent writes, and 15 percent reads.

Examining the workflow of EMC Isilon customers in the EDA industry found that the workload is attribute-intensive because of the large number of files stored in a deep and wide directory structure. To serve I/O requests with the kind of performance that EDA tools require, the storage system should hold the metadata for the working data set in memory or in the highest performance tier.

The input-output profile for an EDA workload looks something like this:

<table>
<thead>
<tr>
<th>METADATA</th>
<th>READ</th>
<th>WRITE</th>
</tr>
</thead>
<tbody>
<tr>
<td>65 percent</td>
<td>15 percent</td>
<td>20 percent</td>
</tr>
</tbody>
</table>

Table 4. Average Metadata Operations for a Typical EDA Workflow

A traditional EDA storage architecture of a controller with disk shelves creates performance bottlenecks for reads, writes, and metadata access. The metadata-intensive aspect of the workload, in particular, saturates a scale-up storage domain—CPU utilization hits 100 percent. The bottleneck produces a latency spike, and users can no longer interact with the storage system in real-time, which can trigger calls to IT support.

In contrast, an Isilon cluster reduces the latency for metadata operations to below a millisecond. Isilon’s distributed architecture eliminates the single-head CPU saturation point of a controller so that you can run a large number of concurrent jobs, which induces a large number of metadata operations, without saturating the storage system.

With its distributed metadata architecture—metadata resides on every node—the OneFS file system separates data from metadata. An Isilon cluster can store all the metadata for the entire data set on SSDs. Storing the data set’s metadata on SSDs is a cost-effective method of eliminating the metadata-access bottleneck because metadata tends to be a mere fraction of the total data set.

With traditional approaches to storage, caching can be useless when there are 500 jobs accessing different data sets. Because the data access patterns for certain aspects of an EDA workflow is random, the cache hit ratio tends to be low, and in such situations the storage system must pull the metadata from disks, increasing the latency of input-output operations.

OneFS goes beyond using flash for the namespace cache. OneFS is in fact not dependent on the cache because OneFS can pin all the metadata for different data sets to SSDs—meaning that the metadata is always resident on flash. As a result, an Isilon cluster can deliver a 100 percent cache hit ratio for all applicable data sets without increasing the input-output latency.

Similarly, while certain technologies may provide the ability to scale out, they still depend on a centralized metadata server while they only scale the data servers. Furthermore, those technologies scale at the file layer, not at the block layer. Given that EDA workload is metadata intensive, the use of a centralized metadata server can become a bottleneck. Whereas other approaches to storage must retrieve metadata from a metadata server and spinning media, OneFS can retrieve all the metadata from SSDs.

**IMPROVEMENT OF RUNTIMES FOR CONCURRENT JOBS:** For an EDA workflow, the Isilon scale-out distributed architecture eliminates CPU bottlenecks and optimizes the performance of processing concurrent jobs. Because all the nodes work in parallel, OneFS automatically distributes jobs with SmartConnect to each node instead of running all the jobs against a single controller or requiring the manual distribution of jobs to controllers. With the right number of nodes, the cluster can process concurrent I/O requests from the compute grid with the optimal level of performance. The distributed, parallel architecture of an Isilon cluster can reduce wall clock runtime for concurrent jobs after the time that other traditional storage systems hit a saturation point for the controller.

One company that uses Isilon to eliminate its storage bottleneck for concurrent jobs is Altera. “Take the largest FPGA we’ve ever built, at a process node that required more simulation than any we’ve ever used before, all on a project time line that was ultra-aggressive and we found only one storage solution that could handle our concurrency demands: Isilon,” Altera’s design automation manager said. With an Isilon cluster, Altera completed over 5 times the number of 20nm simulations in less time than all of the 28nm simulations for the previous generation of its flagship device.

For most workflows, the intensity of I/O operations necessitates solid-state drives (SSDs), write coalescing, coherent caching, and clustered RAM to efficiently serve requests for files and metadata. EMC Isilon S-Series nodes deliver the performance that EDA workflows and workloads demand. A fully scaled EMC Isilon cluster can supply up to 1.1 million NFS SPECsfs2008 file operations per second with more than 100 GB/s of aggregate throughput. The Isilon S200 node combines SSDs with 10,000 RPM 2.5-inch Serial
Attached SCSI (SAS) drive technology, two quad-core Intel CPUs, and up to 13.8 TB of globally coherent cache. SSDs accelerate namespace-intensive metadata operations and access to latency-sensitive data sets. For smaller workloads, the nodes in the Isilon X-Series can also include SSDs to cost-effectively improve performance.

To help eliminate bottlenecks, OneFS includes several cache types, a prefetching option, and dual 10 gigabit Ethernet (10GbE) connections per node. The globally coherent cache provides rapid access to stored data by connecting to any node. You can easily scale the coherent cache to hold your working data set in memory as the data set grows by adding more nodes. The prefetching option can tune OneFS to address streaming, random, or concurrent access patterns. The dual 10 GbE connections to each node support the high levels of network utilization that take place during EDA’s simulation and verification phases.

Finding an Isilon cluster’s optimal point—the point at which it scales in processing concurrent jobs and reduces wall clock runtimes in relation to other systems for the same workload—depends on the size of your compute grid, the number of jobs, the working data sets, and other factors. Isilon recommends that you work with an Isilon representative to determine the number of nodes that will best serve your workflow.

### Storage Efficiency and Utilization

A typical EDA data set consists of about 90 percent or more small files and 10 percent or less large files stored in a file system of thousands of project directories. About 60 percent of the data is active; 40 percent is inactive. Snapshots usually back up the data for short-term retention combined with a long-term backup strategy, which typically includes disk-to-disk backups or tapes.

With traditional filers, an EDA workload utilizes disk space inefficiently. Capacity becomes unevenly distributed across volumes. Of the raw total storage capacity, about 65 percent is usable because of the overhead for parity, spares, snapshot reserve space, the uneven distribution of data, and padding for performance. In practice, overhead often consumes 50 percent of the total capacity because many of the file systems contain pockets of empty space.

An Isilon cluster eliminates much of the overhead that traditional EDA storage systems require. By not having RAID groups, OneFS evenly distributes, or stripes, data among a cluster’s nodes with layout algorithms that maximize storage efficiency and performance. The system continuously reallocates data to conserve space. At the same time, OneFS protects data with forward error correction, or FEC—a highly efficient method of reliably protecting data. The capacity overhead for data protection is about 20 percent in aggregate, not per individual volume.

In practice, an Isilon cluster runs at between 70 and 80 percent efficiency for an EDA data set because it is the large files that dominate utilization, saving as much as 20 to 30 percent of capacity over traditional storage systems. Even when small files make up more than 90 percent of an EDA data set, they consume only 10 percent or less of the capacity. As such, any inefficiencies in storing small files are overshadowed by the efficiencies in storing large files. And as an EDA data set increases in size, an Isilon cluster moves closer to 80 percent efficiency.

For example, to store 100 TB of an EDA data set—say 90 TB of large files and 10 TB of small files—an Isilon cluster requires about 142.5 TB of raw space, as illustrated in the following table:

<table>
<thead>
<tr>
<th></th>
<th>LOGICAL (USED)</th>
<th>PHYSICAL (RAW)</th>
</tr>
</thead>
<tbody>
<tr>
<td>SMALL FILES</td>
<td>10 TB</td>
<td>30 TB</td>
</tr>
<tr>
<td>LARGE FILES</td>
<td>90 TB</td>
<td>112.5 TB</td>
</tr>
<tr>
<td>TOTAL</td>
<td>100 TB</td>
<td>142.5 TB</td>
</tr>
<tr>
<td>EFFICIENCY</td>
<td></td>
<td>70 percent</td>
</tr>
</tbody>
</table>

Table 5. Typical Raw to Usable Ratios for an EDA data set
Furthermore, in the file count vs. capacity distribution scenario shown in the following charts, the level of efficiency is 79 percent:

![File Count Distribution](image1)

![Capacity Distribution](image2)

Figure 3. Large files vs. small files by file count and by capacity.

In addition, you can nondisruptively expand the capacity of the cluster by monitoring the used space on its single volume and then adding nodes just in time to address requirements for more capacity. As a result, EDA system administrators can extract much higher utilization rates from the storage system without risking excessive downtime and without imperiling performance.

Compared with traditional scale-up NAS systems, a scale-out architecture provides a more resilient foundation for data protection and data availability. In its 2013 report titled “Critical Capabilities for Scale-Out File System Storage,” Gartner rated EMC Isilon highest among storage vendors for resiliency—the platform's capabilities for provisioning a high level of system availability and uptime.

Isilon's efficient use of storage space reduces capital expenditures. "CapEx reduction is driven by Isilon's much more efficient use of storage compared with traditional storage," an IDC survey of companies found. "For every $1.00 spent on raw storage, Isilon delivers $0.72 worth of storage in use compared with $0.58 for other traditional storage. One key reason is that the utilization rate of Isilon storage is typically much higher than that of traditional storage. With traditional storage, IT administrators often have to set aside significant discretionary reserves for snapshots and performance headroom so that their storage can continue to operate."

According to the IDC survey, companies using Isilon "were able to drop their related facilities costs by 40%, saving $42 per TB. Facilities savings were driven by the more efficient head architecture, which requires less rack space. Depending on the architecture being replaced, Isilon could use anywhere from 25% to 75% fewer racks."

Project directories that are spread across volumes—there can be anywhere from tens to hundreds of directories in a volume and up to thousands of directories spread across multiple controllers—find a convenient home on an Isilon cluster. OneFS automatically balances data across the entire cluster. As a result, an Isilon cluster can be oversubscribed for EDA storage, and you can expand the cluster nondisruptively to meet expanding storage requirements just in time.

Automated tiering further increases storage efficiency. With other storage systems, moving inactive data to a lower tier can require significant manual effort. With OneFS, a policy can automatically move inactive data to a lower storage tier. As the following chart on inactive files shows, as much as 40 percent of EDA file sets can be inactive for 120 days. Automated tiering with OneFS can save more than a third of the money spent on tier 1 storage, as the chart on inactive files by capacity shows. Snapshots can also reside on a lower tier; there is no need to set aside snapshot reserve space.

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**Data Management**

An Isilon cluster simplifies data management to reduce operating expenses. Isilon’s unique scale-out approach ensures a high return on investment through efficient management while seamlessly scaling to match the ever-increasing storage requirements for EDA project data. Specifically, an EMC Isilon cluster solves many common data management problems for EDA companies by providing capabilities that traditional storage systems lack:

**A SINGLE VOLUME.** An Isilon cluster presents all the nodes in a cluster as a single, global namespace with a central point of management. EDA compute clients access the same data by connecting to any node. No storage silos muddle data management.

**SEAMLESS SCALABILITY.** The Isilon file system dynamically expands up to 20 PB in a single volume and over 100GB/sec of aggregate throughput.

**EFFICIENT DATA MANAGEMENT.** Traditional scale-up architectures can result in costly, manual data migrations among components to balance performance for applications. An Isilon cluster evenly distributes data among nodes or within storage pools without leaving hot spots. The hot spots are inherently eliminated by the file system’s scale-out, distributed architecture. There is no need to manually migrate data. With a single volume, an Isilon cluster requires no manual reconfiguration of client mount points. Service-level agreements become predictable.

An Isilon cluster can combine nodes of different types into a single file system and then segment them into storage pools with different capacity-to-performance ratios. EMC Isilon SmartPools® can, for example, implement file pools that maximize performance for working data sets while cost-effectively storing less active data. File pool policies can identify different data sets by age and then distribute them to a storage tier that matches their performance requirements; for more information, see [EMC Isilon Best Practices for Electronic Design Automation](http://www.emc.com/collateral/analyst-reports/idc-ar-business-benefits-of-scaleout.pdf).

An Isilon cluster’s ease of use and ability to rapidly scale saves time and money. "IT productivity gains reduced the operating cost per TB by 3.5 FTE hours or 48%," IDC found in its survey for general IT system administration. 3 For EDA companies, the operating costs of Isilon scale-out NAS may be even lower.

**TECHNOLOGY STRATEGY.** By scaling multidimensionally to handle the exponential growth of EDA data, an EMC Isilon cluster provides the best of both worlds: capacity and performance. The combination helps you adapt to fluid storage requirements, nondisruptively add capacity and performance in cost-effective increments, and improve wall-clock run times for concurrent jobs.

For EDA, the ratio of CPU, RAM, and disk space depends on the workload—factors that make it difficult to size a storage system before you can measure a future project’s workload. Expanding data sets also makes upfront sizing decisions problematic. Isilon scale-out NAS lends itself perfectly to this scenario: You can increase CPUs, RAM, and disk space by adding nodes to match storage capacity and performance with the demands of a dynamic EDA workload.

Isilon NAS gives you the agility to adapt to changing storage requirements. Instead of purchasing a new fully scaled controller, you can nondisruptively add nodes in cost-effective increments to expand performance, capacity, or both. For instance, if you need to...

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access metadata faster, you can add S-Series nodes, which contain SSDs. Or, if you need to move data from an older project to a lower tier to reduce storage costs, you can add one or more cost-effective NL-Series nodes. You can buy additional storage on an as-needed basis as part of a just-in-time procurement model.

The IDC survey cited earlier found that with Isilon storage, companies not only reduced their storage costs by 37 percent but also increased their agility. “The companies in our study also paid less per TB for their Isilon storage, which, when combined with the efficiency advantages, means that they lowered their storage costs by 37%, saving $1,527 per TB. This advantage means that companies can budget more accurately for their storage requirements and more efficiently scale to meet new or fluctuating demand.”

Isilon’s agility has led many companies in the EDA industry to adopt Isilon clusters, and Isilon’s mature, pure scale-out technology is field-validated for EDA. Along the way, EDA companies have found that they can eliminate a single-vendor strategy. As part of a technology strategy to shorten a chip’s time to market, a correctly sized Isilon cluster can reduce the wall clock runtime for concurrent jobs after the time that other storage systems hit a saturation point for the controller. Being able to run more concurrent jobs without saturating the storage system shortens the time it takes to design and build a chip, which can in turn help improve revenue and profit margins.

**CONCLUSION**

EMC Isilon overcomes the problems that undermine traditional NAS systems by combining the three traditional layers of storage architecture—file system, volume manager, and data protection—into a scale-out NAS cluster with a distributed file system. For electronic design automation, the Isilon scale-out architecture eliminates controller bottlenecks to reduce wall-clock runtimes for concurrent jobs, accelerates metadata operations, improves storage efficiency to lower capital expenditures, centralizes management to reduce operating expenses, and delivers strategic advantages to reduce a chip’s time to market.