

Role of ML and AI in DevOps Transformation

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Abstract: *The stream of DevOps has evolved as a set of practices that help organizations to deliver and operate software applications and services more rapidly, reliably, and efficiently on various infrastructures in different environments. DevOps practices generally include processes such as Continuous Integration, Continuous Delivery, Infrastructure Management, Configuration Management, Monitoring, Governance, Communication, and Collaboration. However, with the dynamically evolving business and technology landscape, the DevOps processes require a transformation. The transformation is driven by several factors like the evolution of cloud computing, increased availability of data, security aspects, growing complexity of applications, and speed and agility required by the digital world. The transformation of DevOps can be greatly enhanced with the help of Artificial Intelligence and Machine Learning concepts helping in addressing business challenges, operational challenges, and technology challenges. Organizations can make use of AI/ML models and techniques to address manual activities like risk management, root cause analysis, monitoring, and alerting. It can also enhance the other aspects of automation with the help of process optimization, maintaining a better sync with the technology stack, and making use of automated bots and virtual assistants. The result is informed decision - making, risk mitigation, and corrective actions. All of these lead to lower costs and better effort utilization. This article dives into the various aspects where AI/ML influences and impacts the DevOps transformation into next - generation solutions.*

Keywords: DevOps, Machine Learning, Artificial Intelligence, DevOps Transformation, Automation, Operational Efficiency

1. The background of DevOps

Traditionally, development teams were responsible for creating the software and deploying it in various environments for the users. An infrastructure team would help in providing support for setting up the required infrastructure in different environments. After the software was deployed and running in the production environment, an operations team would get into the picture. This practice gradually evolved under a single stream known as DevOps. The development teams remained responsible for creating the software. Thereafter, the DevOps team took over the responsibility of preparing the infrastructure for the software in different environments and operating the software in different environments. DevOps's main goal was to enable organizations to deliver software applications and services more rapidly, reliably, and efficiently.

The following key principles were formulated for the scope of DevOps and the success of its goals.

Continuous Integration (CI): Frequently integrating the unit - tested code into a common repository allowing automated testing and code scans.

Continuous Delivery (CD): Frequently building, testing, and promoting the software to various environments repetitively and reliably for the different categories of users, like testers, analysts, and end - users.

Infrastructure Management: Codifying the infrastructure components, such as servers, network, storage, platforms, and databases, and ensuring that it is simple and easy to configure, provision, manage, and monitor those components. It came to be known as Infrastructure as Code (IaC).

Configuration Management (CM): Frequently and reliably making changes to the code, platform, and environment configuration to support the changing needs of the software

and its users. Reproducibility and consistency are the main goals of Configuration Management.

Monitoring and Governance: Continuously monitoring the software, platform, environment, and other infrastructure components for gathering data and generating actionable recommendations. It helps with informed decision - making and course correction.

Communication and Collaboration: Ensuring strong, open, and continuous communication to foster better collaboration between different stakeholders of software, such as the development team, the business users, the vendors, and the management team, for the success of the software operations for several years at a stretch.

Adopting DevOps practices helps organizations achieve faster time - to - market, shorter development cycles, improved software quality, improved reliability, lower costs, and greater scalability leading to enhanced customer satisfaction. It also helps in aligning with the agile ways of the software industry and the rapidly evolving fast - paced technology landscape.

2. The Need for DevOps Transformation

While organizations have rapidly adopted DevOps practices and more maturity has come into the practice, there is a constant need to evolve and grow it further. Simply put, the business and technology landscapes are continuously evolving, causing a ripple effect on other areas of software, including DevOps. The need for transformation is primarily driven by the following factors:

Evolution of Cloud Computing: The concept of cloud computing, which started with shared storage, has grown into every dimension of a software lifecycle and leads to the need for transformation.

The increasing complexity of applications: Applications are becoming more complex with new features spreading across various types of devices and platforms leading to higher interdependencies and added complexity.

The increasing demand for speed and agility: The growing digital world needs a continuous fast - paced highly - reliable means of delivering changes to the software more frequently and efficiently.

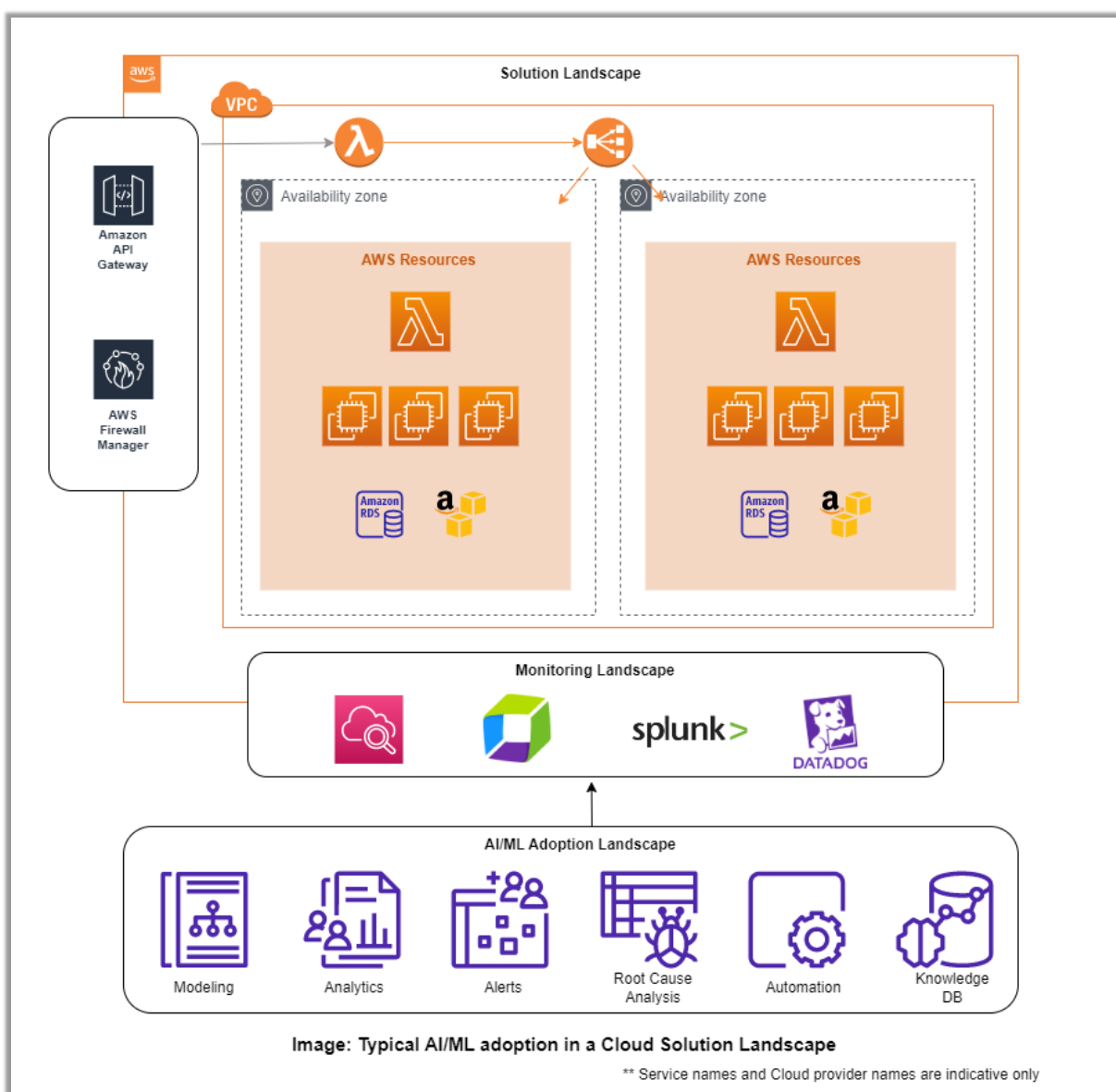
The increasing threat scenarios and security needs: DevOps is evolving into DevSecOps that can seamlessly integrate the security best practices and solutions into the continuous delivery of the software instead of leaving it behind as an afterthought activity.

Large availability of Data: With all the modern monitoring tools and solutions, a huge amount of data is now available that is driving the need for DevOps transformation.

DevOps meets Artificial intelligence

Artificial intelligence is the stream of computing that helps in creating software that can independently learn from the data, logically analyze, and take action. Machine Learning is a field of Artificial Intelligence that allows systems to learn without being explicitly programmed. They feed on large volumes of data from various sources and use multiple dynamic algorithms to identify patterns, make predictions, and provide actionable recommendations.

A typical DevOps solution begins with monitoring of solution landscape and capturing large amounts of data. This data is ingested by the AI/ML models and algorithms to produce actionable insights that can be automated, recorded, or used in other forms by the DevOps teams. We look at a high - level solution architecture with components of DevOps that contribute to AI/ML adoption.



On the path to AI/ML adoption, DevOps solution follows the path of monitoring the operational areas of software, identifying existing and future problems, and proactively

addressing them. However, many manual steps and predefined scripts are involved in present - day DevOps practices.



A look at the way DevOps teams are gearing up to adopt AI and ML in their activities indicates that over 66% of teams are using it for alerts and notifications in 2023. Auto remediation of incidents and Root cause analysis are the other important use cases being used by the DevOps teams.

We have gathered survey results from sources that show the direction in which the DevOps teams are moving forward.

Source: <https://info.opsramp.com/>

It is clear that many problems that DevOps addresses today can be handled in a more optimized and efficient way with the use of Machine Learning and Artificial Intelligence. These problems can be categorized under various heads, like Business challenges, Operational challenges, and Technology challenges. Artificial Intelligence is influencing and improving the solution for each of these solutions.

3. Business Challenges

Managing Risk: Software operations face various risks, such as service outages, unplanned downtimes, cost escalations, vendor lock - in, vendor dependencies, compliance issues, regulatory changes, data breaches, security threats, and more. They can potentially impact the software and its users' causing loss of revenue and penalties. Machine Learning models help in identifying the risks and mitigation plan. It is affecting modern - day risk management in 2 ways: looking at the standard solutions for the solution risks and evolving custom solutions by learning from mitigation plans and actions of other similar projects. It is not possible to manually assess the variety of projects and their associated risks to prepare a foolproof plan, but AI makes it possible.

Decision - making: Software architects are required to analyze the software operations and provide actionable improvements to the application owners for optimizing the cost and time. However, making decisions in such cases is highly opinionated and driven by experience instead of data. Machine Learning models help in looking at the vast stash of log data and other operational metrics and generating trends

that can prove to be a strong basis for decision - making. Additionally, they also help in modeling the outcome of decisions and present a detailed analysis of the impact caused by the decision. It helps the management teams and executives to see the consequences of their decisions and accordingly adjust the course of their actions. A shipping port in the UK needed to upgrade its on - premises servers operating its gate management system, but management did not see the value in the investment. The AI/ML was served with the data, and it showed that a 1 - minute improvement in the performance of the gate would result in over 15% extra business for the port.

Alerts and Notifications: Standard alerts and notifications are part of the application operations. It covers notifications about occurring outages and issues in the system. AI/ML models help in identifying the issues that are not visible in the manual or guided analysis of log data. They look into the data collected over several years, issues faced by the operations teams, and solutions employed. They present insights that can help in preventing high - impact outages in advance. A mere 5 - min slowness in the check - in process of high - speed rail in France would result in long queues at the station counter resulting in greater demand for station staff and increasing the cost of operations. AI/ML would correlate the logs data and notify the need to refresh a server instance even before it hits its configured threshold to maintain the desired levels of performance.

Operational Challenges

Automating Tasks: As part of DevOps practices adoption, several tasks have been automated and sequenced as part of various workflows. Human engineers are still required to make sure the tasks are continuously updated and executed in the correct order for obtaining the desired outcomes. With the help of AI, tasks such as static and dynamic code scans, vulnerability assessments, and penetration tests can be automated using multiple tools. Chaos testing is another potential candidate that AI can manage in a much better way than a manually configured process. The idea is to introduce randomness in the automated processes that can take the simulations and test environments closer to their production counterparts.

Root Cause Analysis: Modern - day DevOps teams spend a significant amount of time unearthing the reasons for the issues faced during the configuration or operation of applications in various environments. AI/ML models employ data analysis and lead to predictive analytics of issues. They perform anomaly detection during the live operations, perform dependency mappings and correlations to the previous problems, and provide operational alerts for potential problems in advance. They go to the extent of troubleshooting the problems automatically without human intervention and avoid the issue altogether. Another important aspect is that the learning is recorded as part of the Knowledge database enhancement that can be referred to for further risk mitigation planning in the future for the same or other solutions.

Optimizing Processes: DevOps teams follow several well - defined processes in executing their responsibilities. Apart from supporting the CI/CD pipelines, Infrastructure

provisioning, and application operations, DevOps teams perform many manual activities. One of the key activities is Incident triage and prioritization. AI/ML models can help in analyzing and categorizing the incident data and reprioritizing the work logs and incident queues for the team. That would save significant time for skilled engineers. Another key activity that can be easily supported by AI/ML model is the capacity planning of the infrastructure and environments. That would help the teams in spending their time, effort, and budget in acquiring and preparing infrastructure easily. AI/ML can also help in preparing strategies for releasing new application versions, managing parallel upgrades, ensuring version compatibility with dependent applications, and mitigating operational risks. Change management is another area where AI/ML models can reference previous variants to propose a thorough strategy allowing one to make informed plans and decisions. A whole lot of manual processes can be made exponentially efficient by the use of AI.

Technology Challenges

Technology Stack: The latest applications vary significantly from each other in the way they are built, released, deployed, and operated. Unlike the olden days of application deployment, where almost every software followed the pattern of using a database server, an application server, and a web server, modern - day software can run independently with or without platform software. Most modern applications come with ready - to - run images that need environment - specific configuration to be injected during the operations. AI/ML can learn the deployment practices and requirements and produce reusable pipelines that can be used with any new application. This would reduce the time of manual analysis and scripting by a DevOps engineer for every newly developed or upgraded application. It takes away the risk associated with the new technology stack and its challenges in the production environment.

Monitoring and Corrective actions: AI can augment monitoring and alerting systems used in DevOps. By employing machine learning algorithms, AI can analyze large volumes of data collected from various sources, such as application logs, metrics, and user feedback, to detect anomalies, predict performance issues, and provide proactive alerts. This helps DevOps teams identify and resolve problems more efficiently, improving system reliability. Regular monitoring also leads to effective capacity planning and capability creation for addressing performance bottlenecks. It helps in predicting and applying scaling and reliability strategies for the applications in the production environment.

Chat Bots and Virtual Assistants: Chatbots improve collaboration and communication amongst and around the DevOps teams. External teams can gain easy access to the knowledge base and actionable solutions for all types of issues, irrespective of whether the issues occur frequently or once in a while. Similarly, AI - powered Virtual Assistants provide instant information to various teams, including management and governance teams about the infrastructure and application operations. They can be the human interface running on top of ML - fueled models, data, and analytics.

4. Conclusion

In summary, ML and AI technologies bring automation, optimization, and intelligent decision - making capabilities to various stages of the DevOps lifecycle, enabling organizations to achieve faster software delivery, improved reliability, and enhanced operational efficiency. With the help of AI/ML models in DevOps, organizations are expected to increase the agility of operations, reduce their operational risks, improve application security, have better decision - making processes based on evidence and simulation models, increase efficiency, and most importantly improve the experience of their staff as well as customers. The end result is lower cost and effort.

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