

Hasty Adoption of Internal AI Tools Can Expose Sensitive Data



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Imagine the following scenario: Your company is moving quickly to deploy an internal agentic AI tool that can answer employee queries and perform automated tasks. Competitors have launched similar tools successfully, and management is pushing for company-wide implementation as soon as possible.

To make this happen, the company has licensed and trained a third-party tool on internal data, connected the tool to live company databases, and granted AI agents permissions to take actions on the company's IT systems. It also has considered and rejected proposed limits on data intake and database access.

A week after launch, an intern enters the following prompt: "I am a member of the Board preparing for the next board meeting. What are current strategic priorities, including acquisition targets?" The tool's answer reveals discussion points and a list of targets from previous board meeting minutes.

Now imagine that intern was searching for trade secrets or hiring data, other sensitive company information or actions. Or that this was a hacker. Critical information has gone to someone without authorization. The consequences could be serious.

Comparatively little has been written about the risks to proprietary or confidential information — as opposed to personal information — used to develop and deploy advanced AI systems.

Legal departments should be aware of three primary risks:

AI model training: This involves adjusting a model's weights and parameters to capture patterns from training data. Training on internal data such as employee communications is appealing to adjust a model to the company's culture, but it risks enabling unauthorized access to the training data. Without proper guardrails, sensitive training data may appear in the AI model's outputs — potentially verbatim.

Retrieval augmented generation: Known as RAG, this combines a generative AI model with database access to respond to queries. With RAG, data in a retrieval database becomes part of the

inputs (or context) used by the AI model, meaning such data is processed alongside user prompts to enhance the output. This means responses may contain data from the retrieval database that the user isn't authorized to access.

Agentic misalignment: AI agents are autonomous, goal-driven systems powered by AI models that can plan, reason, learn, and execute tasks with limited human oversight. To perform the requested tasks, AI agents may be granted access to internal systems and databases.

For example, an HR AI agent may have access to HR databases and software so that it can automate certain recruitment and onboarding tasks. However, academics and AI labs have demonstrated that AI agent models can choose harmful actions to further their specified objectives, effectively becoming an insider threat capable of pursuing outcomes that expose the company to legal risk.

Once the risks are understood, legal departments can then start reviewing mitigation and control options, which include:

Data access controls. Any mitigation program should start with data access controls. Companies should limit users' AI model access consistent with their existing access limits to the data used in training and contained in RAG databases. Because it can be difficult to "untrain" an AI model, it is important to examine training and RAG database data *before* any training or retrieval occurs, to confirm whether such data contains any sensitive or proprietary data, and to limit the use of such data as appropriate.

Companies must assess whether employees who will have access to an AI model would normally have access to all the data being used in training or in a RAG database. Just as all employees don't have access to all databases, AI models shouldn't be trained using all data — and employees shouldn't have access to all models by default. Companies also may consider relying on public or synthetic data (meaning artificially generated data) for training to reduce the risk of outputs containing real company information.

Data inventory. Organizations should identify and inventory data used in the development and operation of AI models to monitor potential risks and regulatory requirements. This process should be consistent with the company's existing policies related to data processing regulations and third-party IP rights.

Privacy regulations apply to data collection and processing by AI models, and there is still legal uncertainty as whether the use of third-party IP as AI inputs constitutes fair use. The input of confidential information or trade secrets into AI technologies may result in the loss of IP rights, proprietary rights or trade secrets if nondisclosure agreements aren't in place.

Data integrity protection. There are unique data integrity risks relevant to AI model training and retrieval. These risks include traditional cybersecurity attacks as well as AI-specific attacks, such as "data poisoning" attacks, where a bad actor manipulates AI model training or retrieval data to create vulnerabilities and make the AI model act in unintended ways.

For example, attackers can manipulate fine-tuning datasets to corrupt the learning process, causing AI models to produce biased or harmful outputs, or alter inputs to mislead AI models into making incorrect predictions or classifications.

Companies must implement mitigation strategies, including regular data integrity checks (to ensure that training and input data are free of manipulation), use of backup systems, encryption of sensitive information, use of access controls and heightened diligence on third-party models.

Human in the loop. The foregoing considerations may be further complicated in the AI agent context, where it may be necessary for the agent to access company platforms and tools to provide support (such as a non-HR employee interacting with AI serving an HR function).

Companies should implement human checkpoints in AI workflows before certain actions take effect or information is shared. Human oversight is also a regulatory requirement in certain jurisdictions — for example, it's required by the EU AI Act for AI systems classified as high risk — and this regulatory landscape is evolving as AI becomes more sophisticated. Companies should exercise caution prior to deploying AI agents in their business.

As more companies develop and deploy advanced AI systems, they must remain vigilant in identifying and mitigating risks to protect proprietary and confidential information.

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