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Monitor data and model quality

PDF (/pdfs/sagemaker/latest/dg/sagemaker-dg.pdf#model-monitor)

RSS (amazon-sagemaker-release-notes.rss)

Amazon SageMaker Model Monitor monitors the quality of Amazon SageMaker machine learning models in production. You can set up continuous monitoring with a real-time endpoint (or a batch transform job that runs regularly), or on-schedule monitoring for asynchronous batch transform jobs. With Model Monitor, you can set alerts that notify you when there are deviations in the model quality. Early and proactive detection of these deviations enables you to take corrective actions, such as retraining models, auditing upstream systems, or fixing quality issues without having to monitor models manually or build additional tooling. You can use Model Monitor prebuilt monitoring capabilities that do not require coding. You also have the flexibility to monitor models by coding to provide custom analysis.

Model Monitor provides the following types of monitoring:

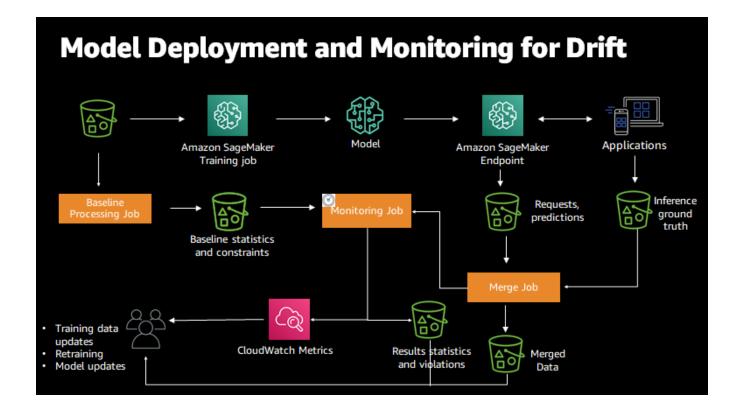
- Monitor data quality (./model-monitor-data-quality.html) Monitor drift in data quality.
- Monitor model quality (./model-monitor-model-quality.html) Monitor drift in model quality metrics, such as accuracy.
- Monitor Bias Drift for Models in Production (./clarify-model-monitor-bias-drift.html) Monitor bias in your model's predictions.
- Monitor Feature Attribution Drift for Models in Production (./clarify-model-monitor-feature-attribution-drift.html) Monitor drift in feature attribution.

Topics

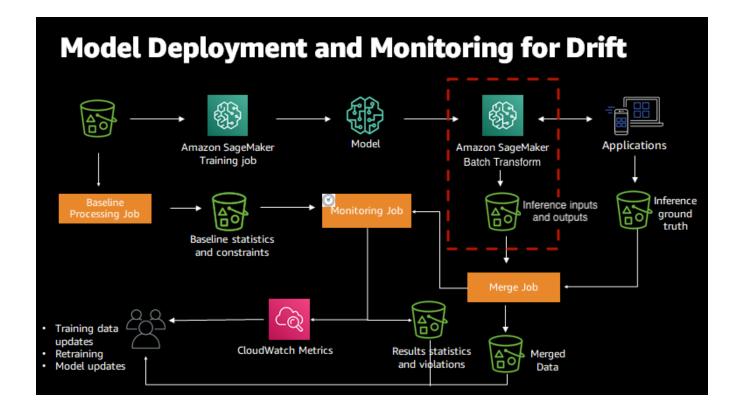
- Monitoring a Model in Production (./how-it-works-model-monitor.html)
- How Model Monitor Works (#model-monitor-how-it-works)
- Capture data (./model-monitor-data-capture.html)
- Monitor data quality (./model-monitor-data-quality.html)
- Monitor model quality (./model-monitor-model-quality.html)
- Monitor Bias Drift for Models in Production (./clarify-model-monitor-bias-drift.html)
- Monitor Feature Attribution Drift for Models in Production (./clarify-model-monitor-feature-attribution-drift.html)
- Schedule monitoring jobs (./model-monitor-scheduling.html)
- Amazon SageMaker Model Monitor prebuilt container (./model-monitor-pre-built-container.html)
- Interpret results (./model-monitor-interpreting-results.html)
- Visualize results for real-time endpoints in Amazon SageMaker Studio (./model-monitor-interpreting-visualize-results.html)
- Advanced topics (./model-monitor-advanced-topics.html)
- Model Monitor FAQs (./model-monitor-faqs.html)

How Model Monitor Works

Amazon SageMaker Model Monitor automatically monitors machine learning (ML) models in production and notifies you when quality issues arise. Model Monitor uses rules to detect drift in your models and alerts you when it happens. The following figure shows how this process works in the case that your model is deployed to a real-time endpoint.



You can also use Model Monitor to monitor a batch transform job instead of a real-time endpoint. In this case, instead of receiving requests to an endpoint and tracking the predictions, Model Monitor will monitor inference inputs and outputs. The following figure diagrams the process of monitoring a batch transform job.



To enable model monitoring, you take the following steps, which follow the path of the data through the various data collection, monitoring, and analysis processes:

- For a real-time endpoint, enable the endpoint to capture data from incoming requests to a trained ML model and the resulting model predictions.
- For a batch transform job, enable data capture of the batch transform inputs and outputs.
- Create a baseline from the dataset that was used to train the model. The baseline computes metrics and suggests constraints for the metrics. Real-time or batch predictions from your model are compared to the constraints, and are reported as violations if they are outside the constrained values.
- Create a monitoring schedule specifying what data to collect, how often to collect it, how to analyze it,

and which reports to produce.

• Inspect the reports, which compare the latest data with the baseline, and watch for any violations reported and for metrics and notifications from Amazon CloudWatch.

Notes

- Model Monitor computes model metrics and statistics on tabular data only. For example, an
 image classification model that takes images as input and outputs a label based on that
 image can still be monitored. Model Monitor would be able to calculate metrics and statistics
 for the output, not the input.
- Model Monitor currently supports only endpoints that host a single model and does not support monitoring multi-model endpoints. For information on using multi-model endpoints, see Host multiple models in one container behind one endpoint (./multi-model-endpoints.html).
- Model Monitor supports monitoring inference pipelines, but capturing and analyzing data is done for the entire pipeline, not for individual containers in the pipeline.
- To prevent impact to inference requests, Data Capture stops capturing requests at high levels of disk usage. It is recommended you keep your disk utilization below 75% in order to ensure data capture continues capturing requests.
- If you launch SageMaker Studio in a custom Amazon VPC, you need to create VPC endpoints to enable Model Monitor to communicate with Amazon S3 and CloudWatch. For information about VPC endpoints, see VPC endpoints (https://docs.aws.amazon.com/vpc/latest/userguide/vpc-endpoints.html) in the Amazon Virtual Private Cloud User Guide. For information about launching SageMaker Studio in a custom VPC, see Connect SageMaker Studio Classic Notebooks in a VPC to External Resources (./studio-notebooks-and-internet-access.html).

Model Monitor Sample Notebooks

For a sample notebook that takes you through the full end-to-end workflow using Model Monitor with your real-time endpoint, see Introduction to Amazon SageMaker Model Monitor (https://sagemaker-examples.readthedocs.io/en/latest/sagemaker_model_monitor/introduction/SageMaker-ModelMonitoring.html).

For a sample notebook that visualizes the statistics.json file for a selected execution in a monitoring schedule, see the short of visualization of the work of the monitoring of the statistics of the statistics of the second of the statistics of the statistics of the statistics.