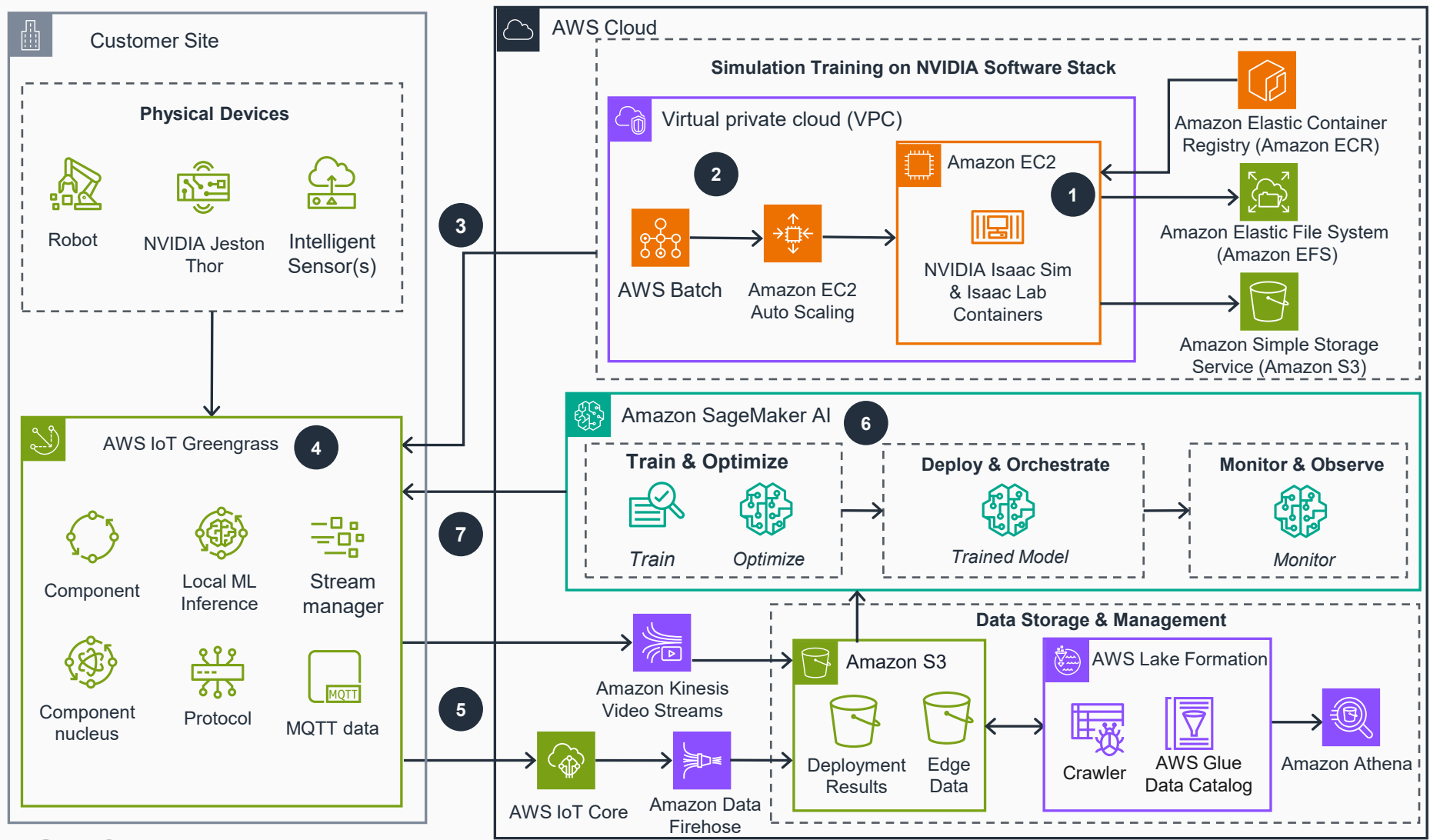


Guidance for Physical AI for Robotic Applications on AWS

This architecture enables autonomous robotics application development through reinforcement learning from physics-based closed-loop feedback systems. Models trained in simulation are deployed to robot controllers for real-world testing. Edge devices collect multi-sensor data that flows to the AWS cloud for continuous model refinement and redeployment. Customers can choose to start with either approach and iterate continuously. This slide shows steps 1-6.

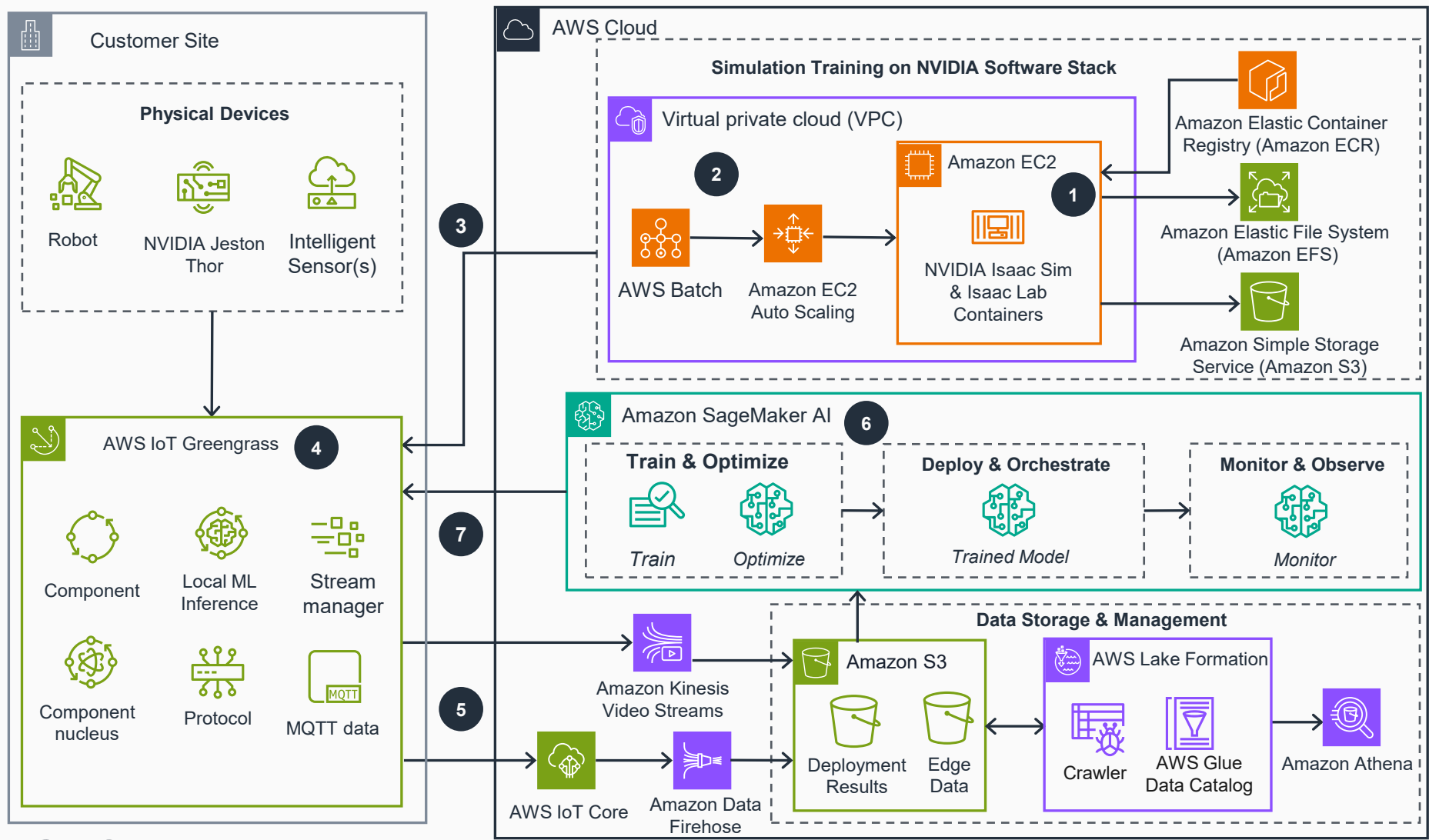


1. Begin Simulation training with robotics reinforcement learning using NVIDIA Isaac Sim containers deployed on GPU-powered **Amazon Elastic Compute Cloud (Amazon EC2)** instances for modeling and NVIDIA Isaac Lab to scale training scenarios. Test physics constraints and scenarios through multiple iterations within this simulation loop.
2. **AWS Batch** orchestrates simulation workloads across GPU-powered **Amazon EC2** Auto Scaling groups to dynamically scale compute resources based on demand.
3. One-way Deployment: The trained ML model with robot policies are deployed one-way to **AWS IoT Greengrass** running on physical controllers that interface with robots at the edge.
4. **AWS IoT Greengrass** components process real-time physics and environmental feedback data from sensors including cameras, audio, gyroscopes, force, accelerometers, contact sensors, joint encoders, position, and pressure sensors.
5. **AWS IoT Greengrass** sends MQTT sensor data through **AWS IoT Core** and **Amazon Data Firehose** to **Amazon Simple Storage Service (Amazon S3)** data lakes, while video streams flow via **Amazon Kinesis Video Streams** to **Amazon S3** for storage and management
6. **Amazon SageMaker AI** processes batches of real-world data to train and/or retrain and optimize models, bridging sim-to-real gaps between NVIDIA Isaac Sim virtual simulation and actual robot operations.



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7

Continuous Deploy and Monitor: Refined ML models trained in **Amazon SageMaker AI** are deployed to **AWS IoT Greengrass** on the robot edge. Inference is performed using these models to optimize robot behavior and meet performance goals. A monitoring layer tracks metrics, detects drift, and triggers retraining. ML model iteration continues through this cycle: robots generate operational data, models are refined based on real-world performance, and improved models are redeployed.

