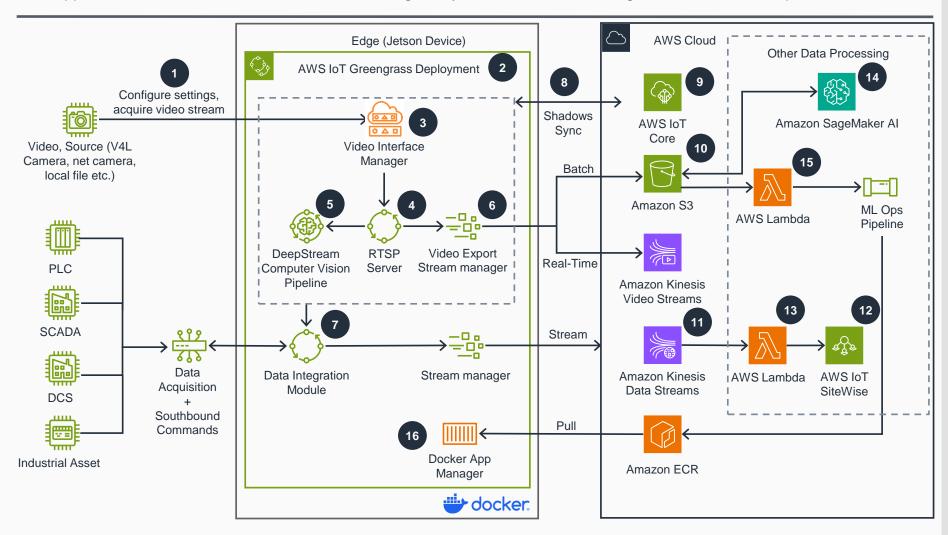
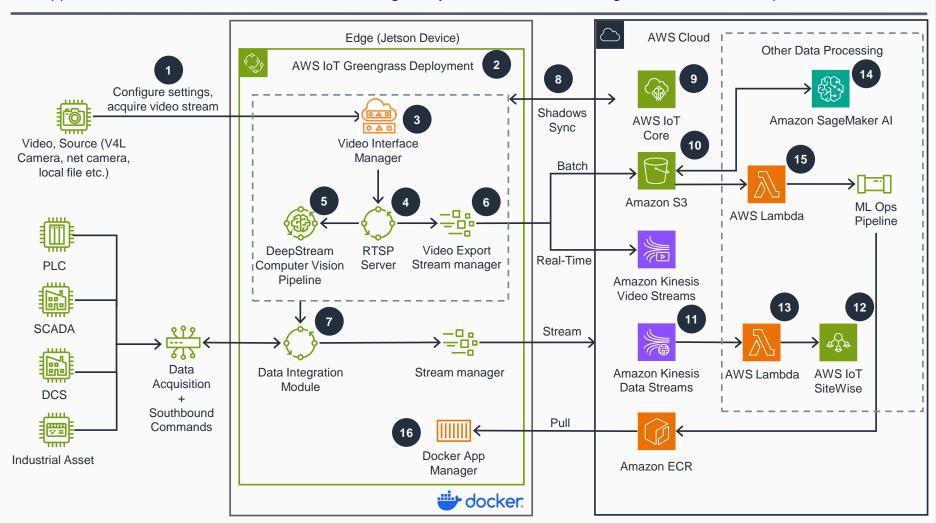
This architecture diagram shows how the EdgeInsight solution accelerator enables rapid deployment of AI-on-the-edge computer vision applications for near real-time industrial monitoring, analysis, and decision-making. This slide shows steps 1-5.



- Connect to video feeds from factorymounted cameras or Video Management
 Systems (VMS) for primary data input.
 Supplement training with synthetic data to
 expand test datasets and simulate rare
 defect patterns beyond typical factory
 scenarios.
- Deploy AWS IoT Greengrass on an NVIDIA Jetson device or Industrial PC (IPC) with discrete GPU to run the edge data processing pipeline.
- Create a Data Processing pipeline with Video Interface Manager to standardize interactions with Open Network Video Interface Forum (ONVIF) cameras, Real-Time Streaming Protocol (RTSP) streams, and GigE Vision systems. This abstraction layer eliminates the need to handle proprietary hardware protocols or vendor-specific specifications.
- Implement an RTSP server as a video stream broker to manage complex routing of video streams. This enables multiple components to access and process video data simultaneously while maintaining system performance.
- Deploy the NVIDIA DeepStream
 Computer Vision (DSCV) pipeline to
 analyze video in near real-time. This highthroughput, low-latency platform uses
 hardware acceleration to perform tasks
 from object detection to quality inspection.

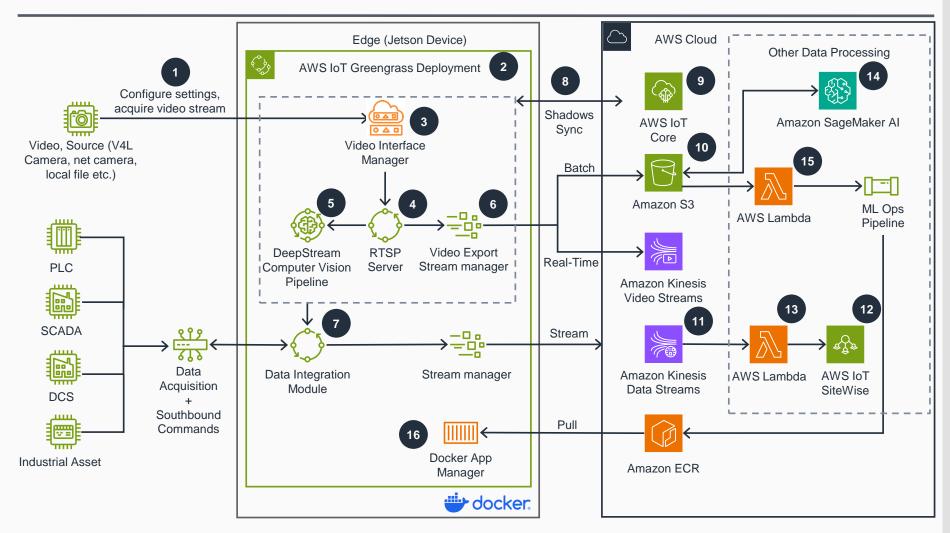
This architecture diagram shows how the EdgeInsight solution accelerator enables rapid deployment of AI-on-the-edge computer vision applications for near real-time industrial monitoring, analysis, and decision-making. This slide shows steps 6-10.



- Use the Video Stream Export Manager (VSEM) to transfer processed or raw video to AWS Cloud storage and services through batch or near real-time operations.
- Configure the Data Integration Module to process inference data and execute business logic at the edge. Transform raw DeepStream CV data through filtering, aggregation, and enrichment. Combine video analytics with OT process data from Programmable Logic Controllers (PLC), Distributed Control System (DCS), and Supervisory Control and Data Acquisition (SCADA) systems for comprehensive context, such as correlating conveyor state data with object detection.
- Implement near real-time and batch video/data stream processing through interconnected services for seamless edge-to-cloud operations using AWS IoT Core, Amazon Simple Storage Service (Amazon S3), Amazon Kinesis Video Streams, Amazon Kinesis Data Streams, and AWS IoT SiteWise.
- 9 Set up **AWS IoT Core** for secure bidirectional communication with edge compute devices. Use device shadows to maintain configuration synchronization during intermittent connectivity or partially air-gapped factory operations.
- Use **Amazon S3** to store video artifacts and ML models for edge deployment, providing scalable object storage.



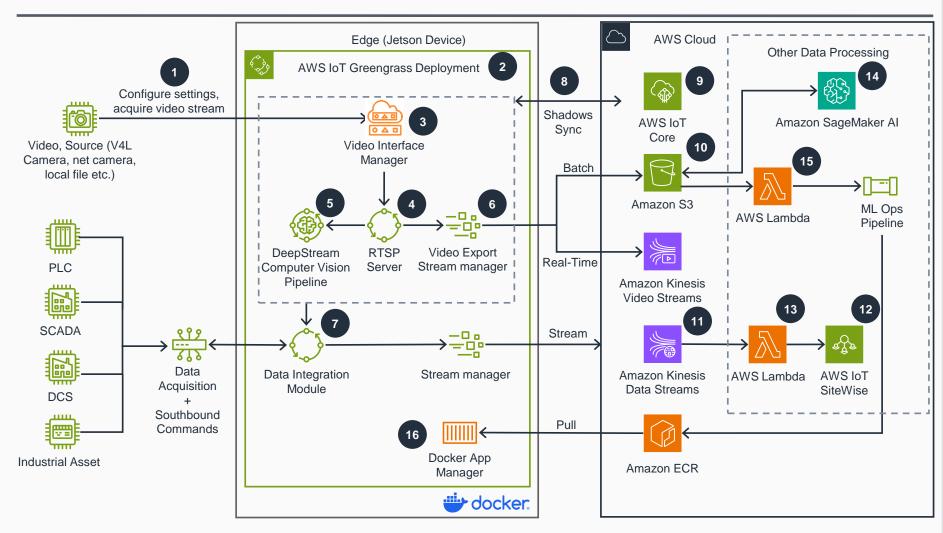
This architecture diagram shows how the EdgeInsight solution accelerator enables rapid deployment of AI-on-the-edge computer vision applications for near real-time industrial monitoring, analysis, and decision-making. This slide shows steps 11-15.



- Use Amazon Kinesis Data Streams and Amazon Kinesis Video Streams to ingest real-time video and inference metadata.
- Configure AWS IoT SiteWise to create digital representations of industrial data hierarchies. Enable production managers to correlate defect rates with machine parameters, linking visual defects to specific temperature, vibration, or other operational fluctuations.
- Use AWS Lambda as a connective layer between services to process edgegenerated data and cloud operations.

 Transform data between services, extract video metadata, process inference outputs, trigger container deployments, and manage over-the-air update notifications and verifications.
- Deploy Amazon SageMaker AI as the ML foundation for EdgeInsight's AI-on-the-Edge computer vision continuous improvement cycle. Data scientists use Amazon SageMaker AI to train models using real or NVIDIA Omniversegenerated synthetic data, then store trained models in Amazon S3.
- Trigger ML Ops pipeline using AWS
 Lambda to package ML models as
 Docker containers and store them in
 Amazon Elastic Container Registry
 (Amazon ECR).

This architecture diagram shows how the EdgeInsight solution accelerator enables rapid deployment of AI-on-the-edge computer vision applications for near real-time industrial monitoring, analysis, and decision-making. This slide shows step 16.



Configure Docker Application Manager to pull images from **Amazon ECR** and deploy software updates and optimized **Amazon SageMaker AI** MLOps models to edge devices, ensuring secure and efficient delivery of improvements.

