tinyML. Summit

Miniature dreams can come true...

March 28-30, 2022 | San Francisco Bay Area



XICCFC



Suitability of TinyML for predictive maintenance in high-tech manufacturing

Prepared for TinyML Summit March 29th, 2022



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Introduction to SensiML

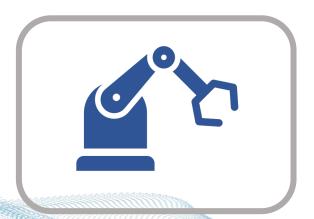
We Sell: TinyML toolkit for Smart Sensors

To: Application developers and system integrators

To enable: rapid development of novel applications for edge devices







SensiML Analytics Toolkit - Market Leading AutoML Technology for IoT Endpoint Algorithms



Introduction to Adapdix

We Enable

Next-level Operational Efficiency & Cost Savings

45% Avg. Maintenance Savings

50% Avg. UDT Reduction

33% Avg. Cycle Time Savings

20% Avg. UPH Increase

Software For

Autonomous Operations

100's Exabytes of Data Ingested

2000+ Models Deployed

100,000+ Anomalies Resolved

100's of Machines Optimized

Benefitting

Global Leaders

2 of the largest, global Semiconductor Manufacturers

Top 10 Semiconductor OEMs

Fortune 500 Communication Companies

Adapdix Enables Autonomous Systems

High-Tech Manufacturing Predictive Maintenance

Case Study

A leading semiconductor manufacture to add predictive maintenance to a photonic optical alignment manufacturing process

The Challenge

- Unplanned downtime through unexpected equipment or process failure drives high cost and low OEE
- Identifying failure root causes in complex system is challenging without high visibility into all the processes

Implemented multistep process to prevent downtime and improve yield

- Identify issues with the current process leading to low yield
- Identify indications of equipment failure to prevent downtime
- Improved yield by increasing the throughput units/hour
- Once the manufacturing process is finalized, identify deviations/anomalies in behavior

The final solution was successfully deployed on over 100 machines in the factory

Successfully Increased throughput by over 50%



Industrial AI: Advanced Analytics at Scale

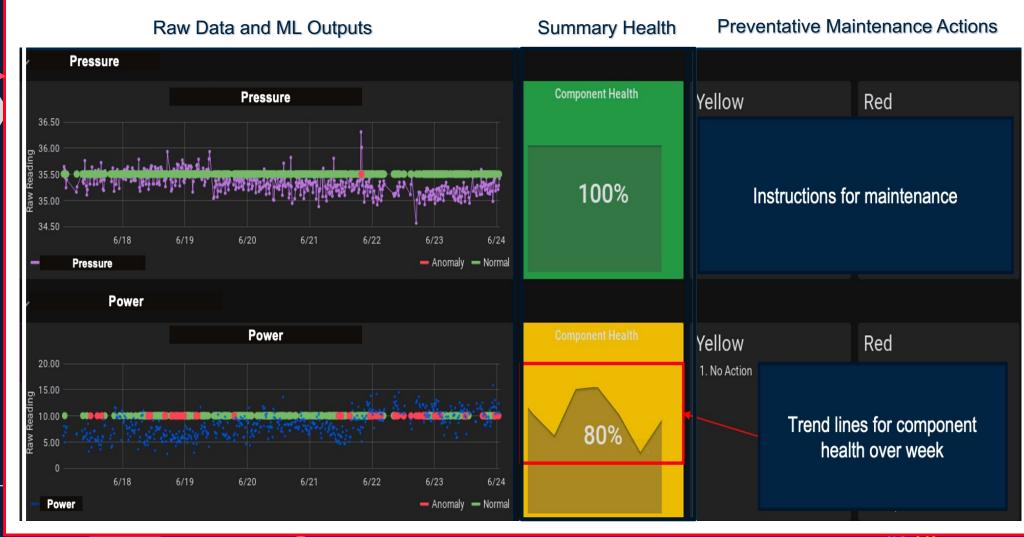
Complex Integration, Data and Communications



Inputs/Outputs

- Sequencing
- Timing
- Latency
- Control
- Memory
- Computational order
- Anomaly detection
- Programmable logic

Synchronization





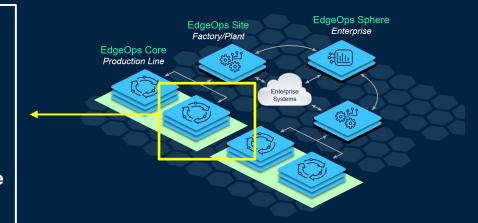
Where Does TinyML Fits into the Solution?

- Using Adapdix EdgeOps Autonomous Systems Platform to orchestrate data flow, analytics, monitoring and automate the tool workflow
- Used SensiML to build TinyML models for real-time time-series classification and anomaly identification

TinyML is only a part of the solution

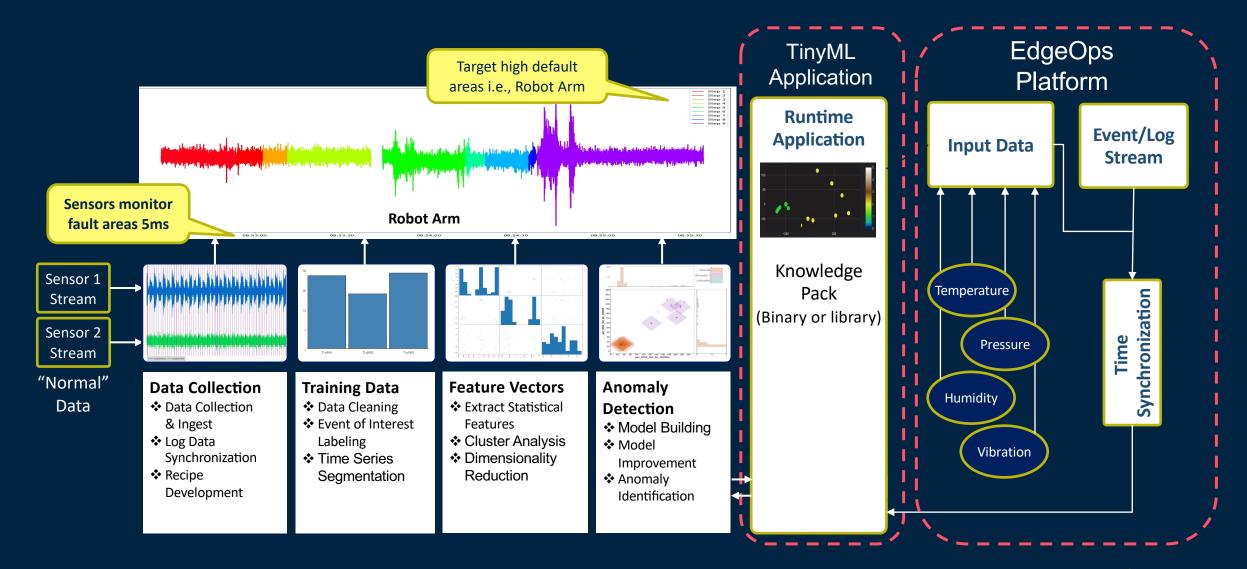
Makes deploying models easier and provides flexibility in how close ML models can run to the sensor data

Allows you to run multiple models in commodity hardware in real-time



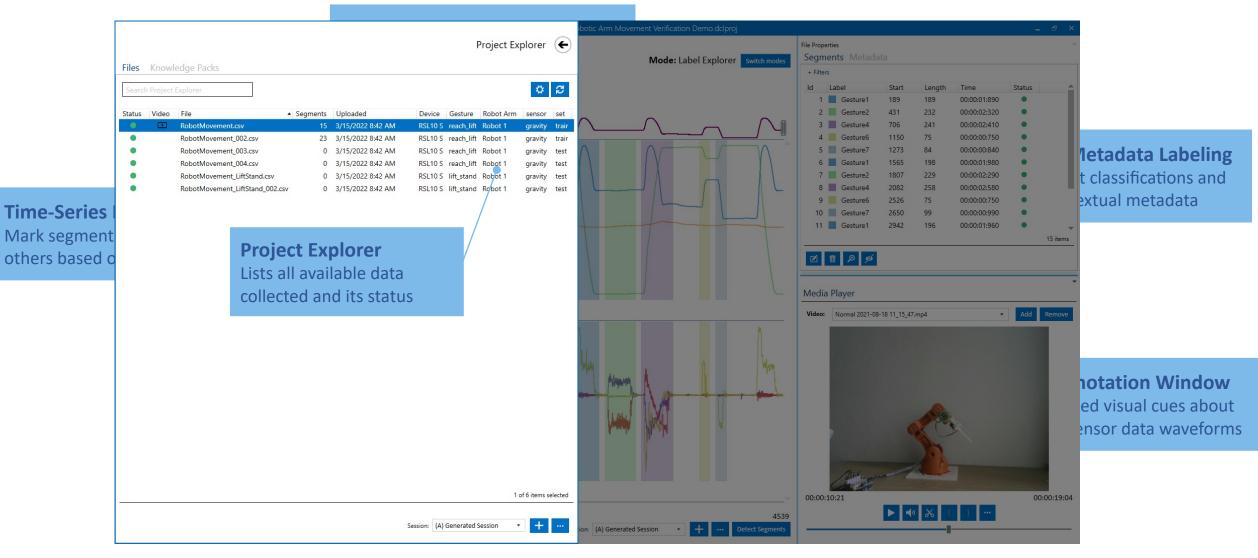


Architecture & TinyML Integration





SensiML Data Capture Lab: Data Curation Made Simple





SensiML Analytics Studio: Efficient AutoML for Edge Sensing

Query Labeled Data

Easily Filter based on metadata

AutoML for Edge Al

Select Feature, algorithm parameters, Tune max model size

Model Visualization

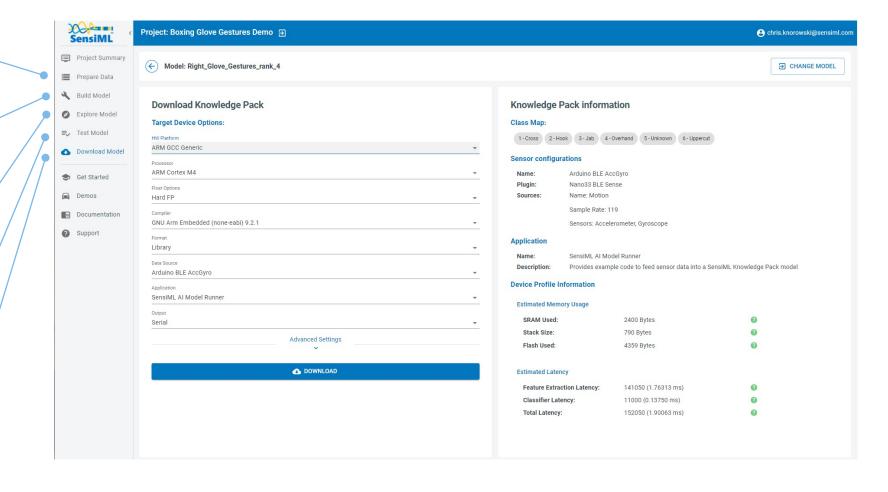
Explore model parameters, visualize features, and model metrics

Test Model

Test Model and validate offline

Download Model

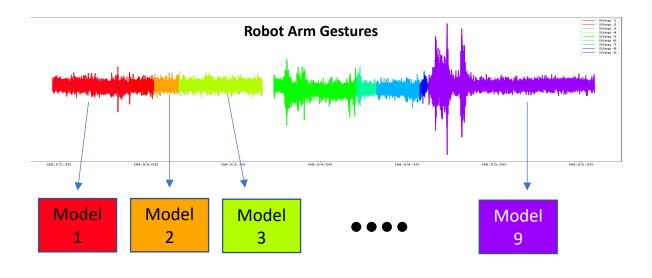
Download Library, Binary or Source code optimized for target device





SensiML Python SDK

- Programmatic access to model building
- Python API for data visualization and manipulation



Pipelines %

A pipeline is a container for a series of data processing steps. The pipeline object allows you to get an existing pipeline or create a new one with a given name. With this object, you can set input data sources, add transforms, feature generators, feature selectors, feature transforms and classifiers. An example pipeline is shown below.

Examples:

Multi Model Knowledge Pack

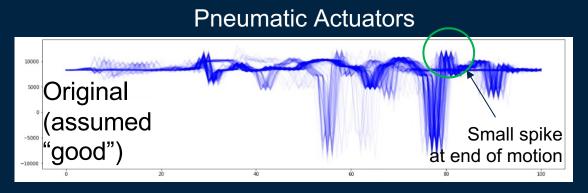
If you want to download multiple Knowledge Packs at once you will need to use the programmatic interface. Models can either be Parent or Children Models. Parent models require a source. Children models require a segmenter_from and parent flag. Optional flags are results which allows a Child model to be called depending on the output of a Parent model.

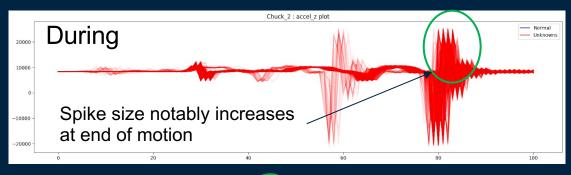
The model graphs are defined in the kb_description. The format for multiple parent models is as follows:

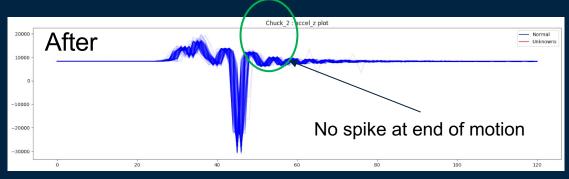
```
{
  "MODEL_1": {
      "source": "<CAPTURE CONFIG UUID>",
      "uuid": "<Model UUID>"
},
  "MODEL_2": {
      "source": "<CAPTURE CONFIG UUID>",
      "uuid": "<Model UUID>"
},
  "MODEL_3": {
      "source": "<CAPTURE CONFIG UUID>",
      "uuid": "<Model UUID>"
}
}
```



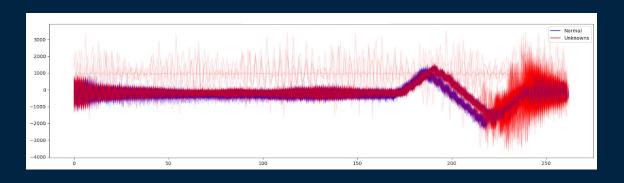
Preventative Maintenance Around Equipment Motion

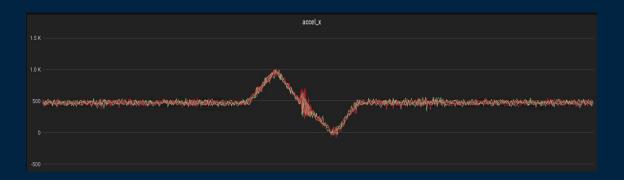






Linear Motor Stage



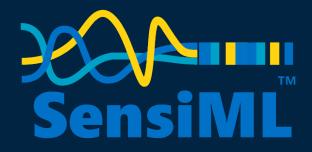




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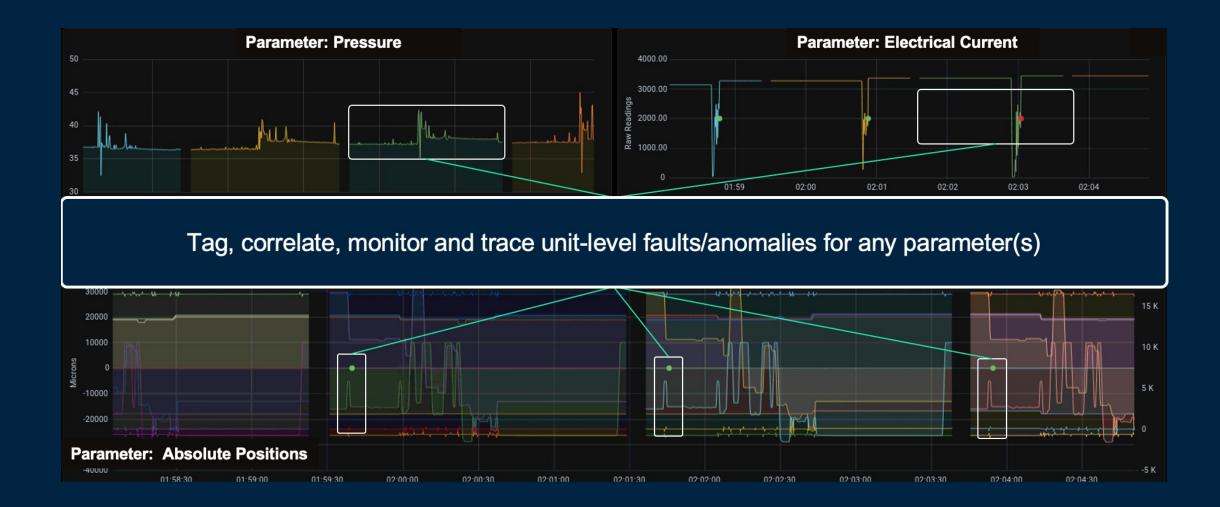
support@sensiml.com https://sensiml.com

Chris Knorowski chris.knorowski@sensiml.com

Additional Detail



Unit-level Fault Traceability





Vibration Use Cases

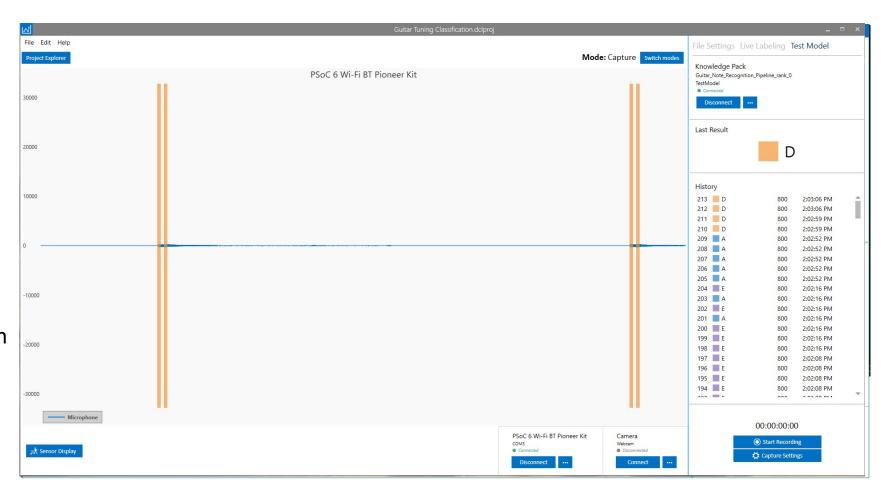






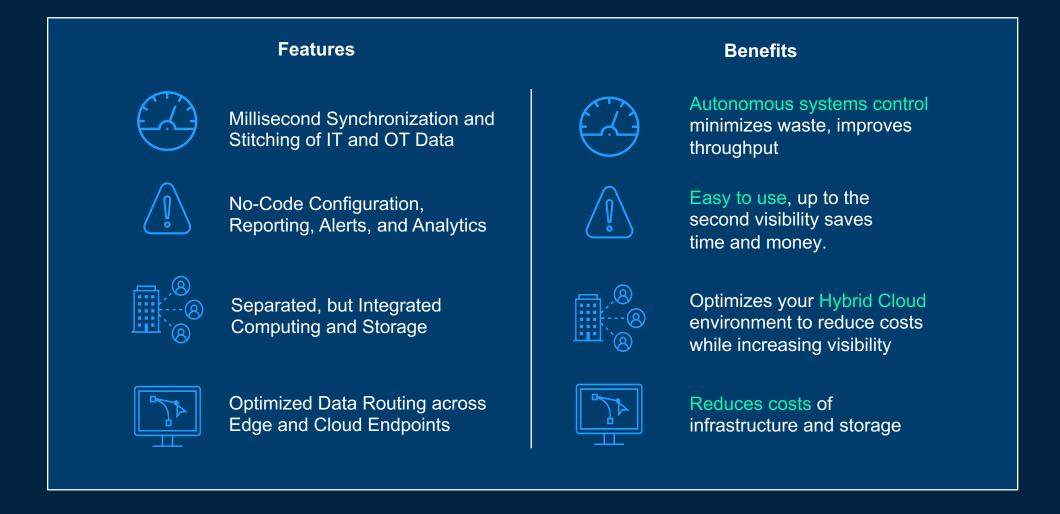
SensiML Data Capture Lab: Data Collection and Validation

- Capture Sensor Data:
 - WiFi/TCP/IP
 - BLE
 - USB Serial Port
 - Import CSV, WAV
- Easily Collect Metadata
- Programmable File Naming
- Live Labeling
- Real Time Model Validation
- Video Capture/Syncrhonization





EdgeOps Platform: Key Features & Benefits





Equipment Improvement Area

Improvement approach:

- Model Power readings
- Alert if anomalies exceed 15%

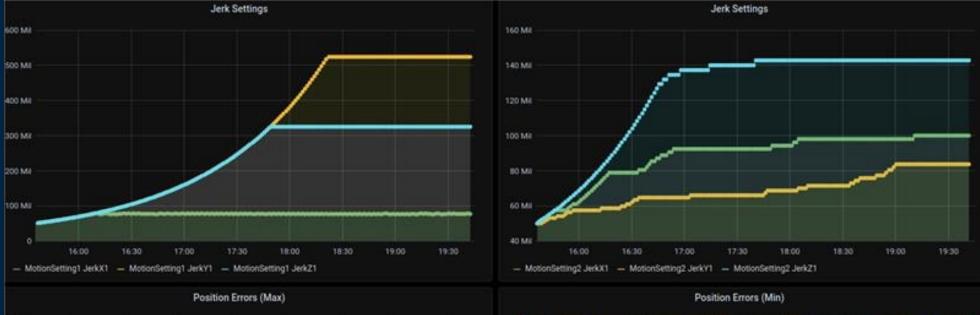
Benefit:

 Decrease downtime due to Low Power



Adaptive Controls

EdgeOps Feedback



Equipment Changes





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