

# NanoLambda : FaaS at All Resource scales for IoT

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# Background

- IoT devices are increasingly prevalent producers of data
- Writing code for processing data on IoT devices remains a challenge.
  - Difficult to implement
  - Portability, security and maintainability all are challenges
- Cloud / Edge processing:
  - Easy to program
  - Easy to scale – FaaS
  - Drawback – Cost of sending data, latency etc.

# Background

- Function as a service –
  - Easy scalability
  - Event driven, Resource conserving
  - Faster development
- FaaS is limited to linux based resource rich systems.
- This paper tries to provide uniform FaaS environment for highly resource constrained IoT devices.

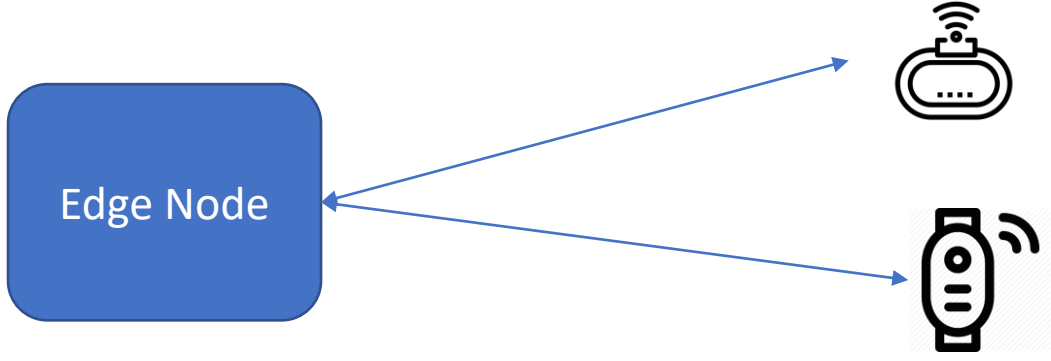
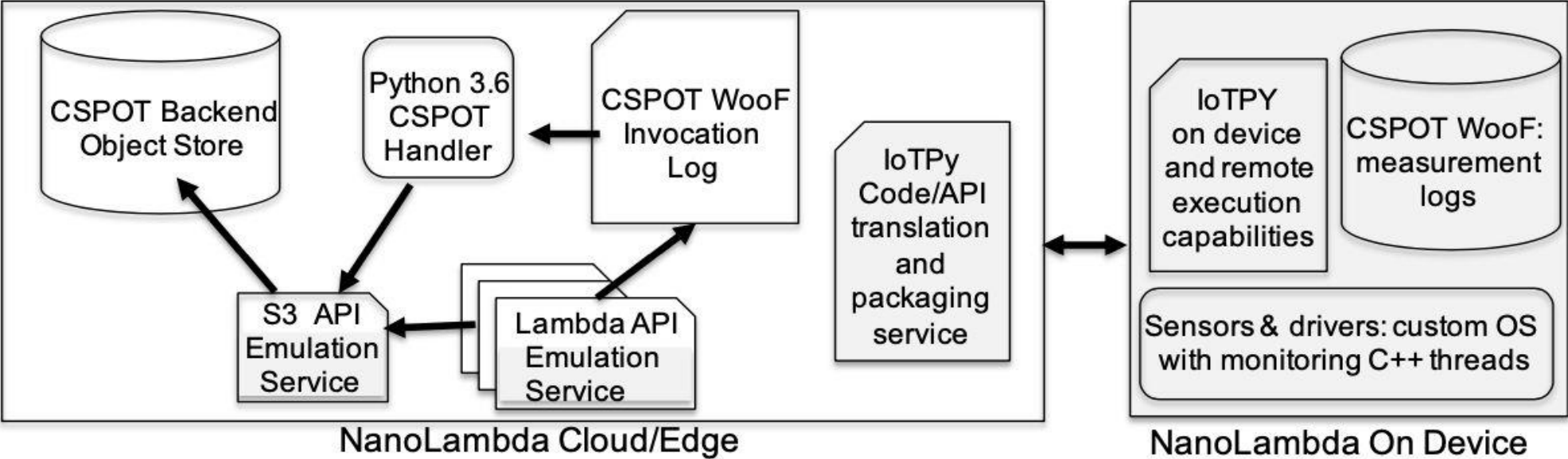
# NanoLambda

- NanoLambda: platform for running FaaS handlers across all tiers
  - On Device
  - Cloud / Edge
  - Compatible with AWS Lambda
- Goals :
  - Ease of development
  - Portability
  - Small code and memory footprint
  - Security (Isolation)
  - Uniform programming methodology
- At the smallest scales
  - ESP8266 with 96KB of ram and 512KB of program flash storage
  - CC3220SF with 256KB of ram and 1MB of program flash storage

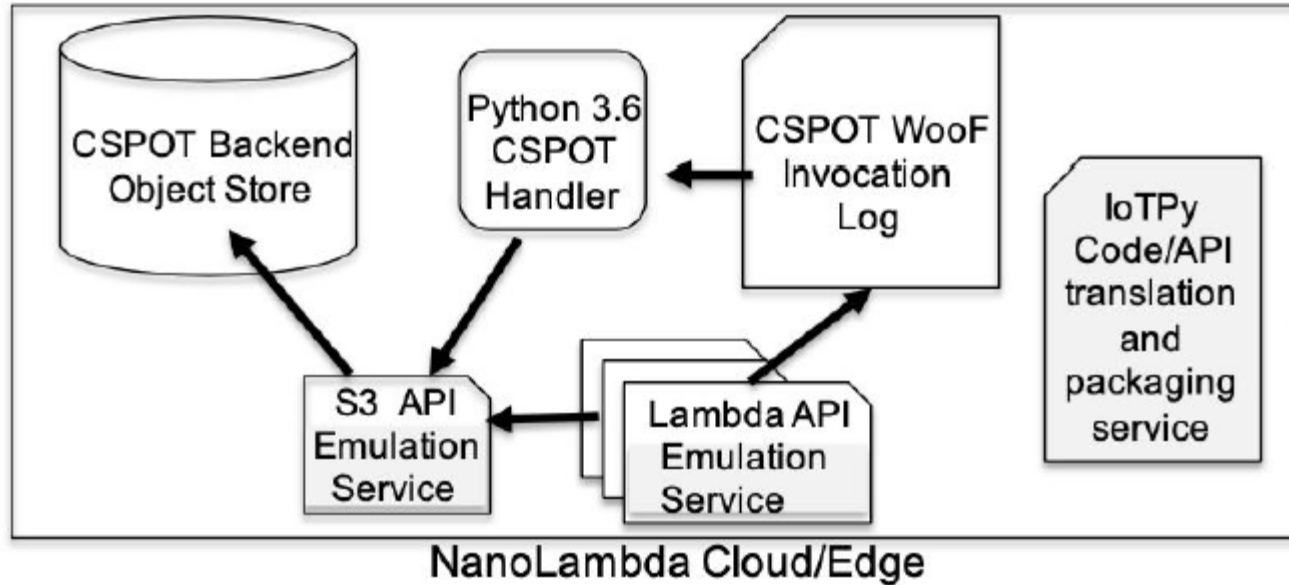
# NanoLambda Architecture

- Based on CSPOT.
  - FaaS deployment model
  - Fault resilient distributed storage
  - Cloud and Edge portability (Only for Linux based systems)
- Comprises of 2 cores systems:
  - NanoLambda Cloud/Edge
  - NanoLambda On Device

# NanoLambda Architecture

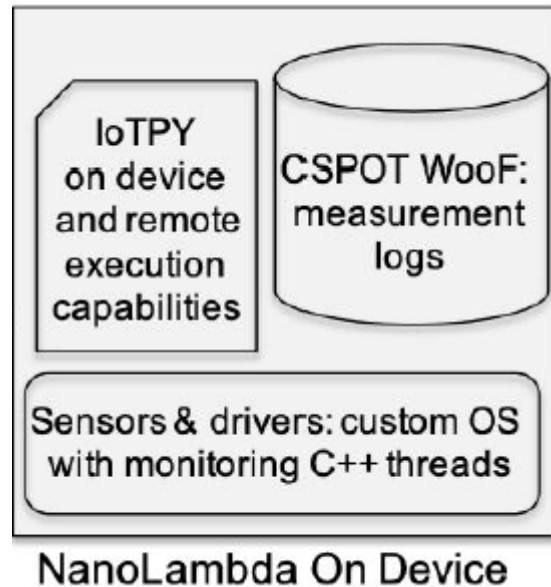


# NanoLambda Cloud/Edge



- Service provides two REST API servers offering
  - S3 emulation
  - A FaaS service that deploys functions written for AWS Lambda
- Built with CSPOT
  - S3 is implemented as a layer on top of CSPOT's append only object storage
  - Lambda is implemented with event handlers triggered by invocation log updates
- Handlers are run in Linux containers allowing for concurrent but isolated execution
- Provides a registry of function definitions stored in S3 service
- Binary API for fetching compiled function bytecode

# NanoLambda on Device



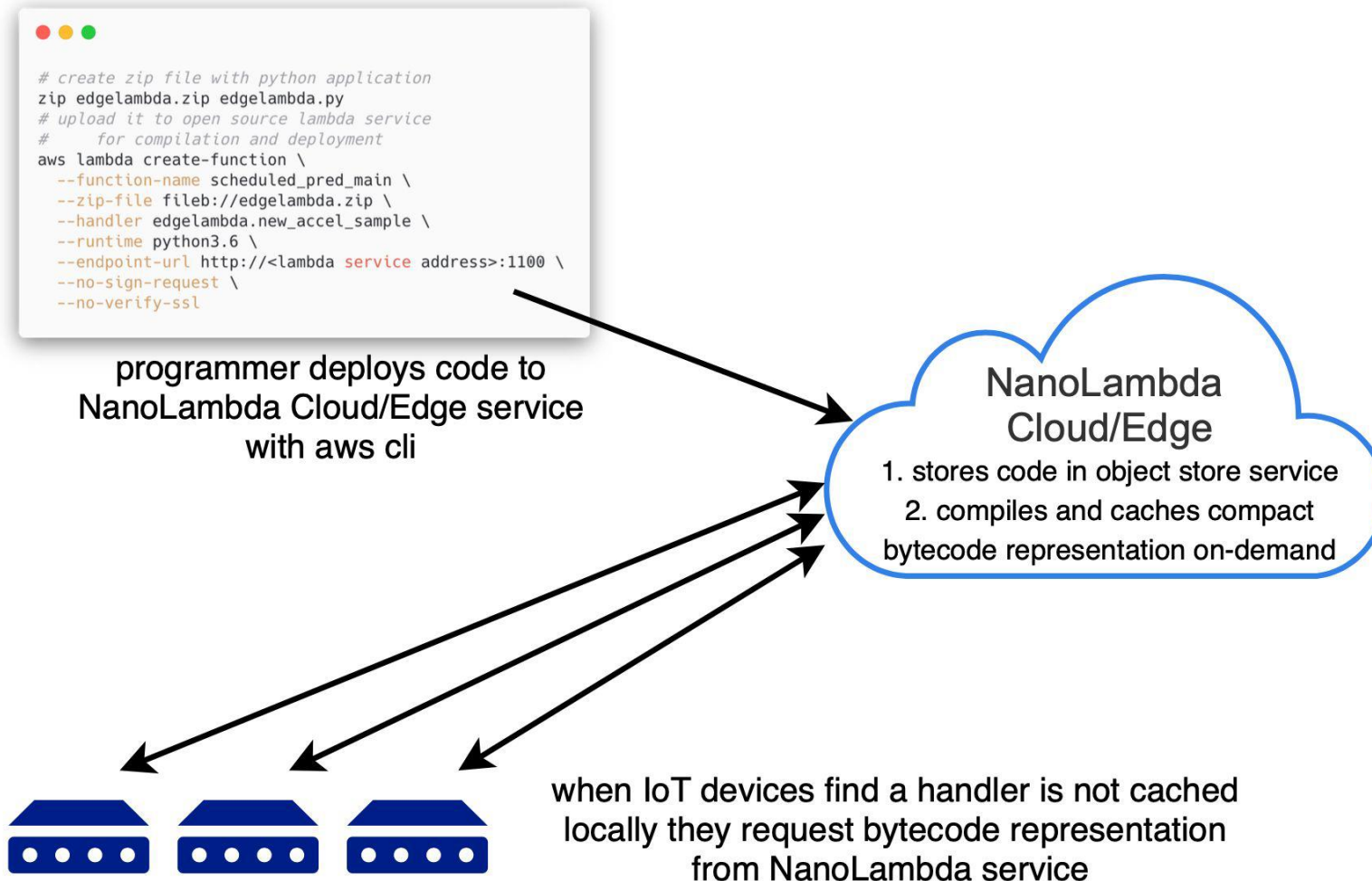
- Runs python handlers on non-Linux IoT devices
- Much like NanoLambda Cloud/Edge, invocations triggered by log events
  - Events can originate from sensors on device
  - Data can also be delivered remotely over CSPOT's network
- Each append runs a C-language handler function invoking IoTpy
- On cold start function bytecode is requested from NanoLambda Cloud/Edge
- IoTpy caches bytecode & interpreter state to accelerate future runs



# IoTPy

- Extremely lightweight Python interpreter built from the ground up with embedding in mind.
- Why not an existing interpreter like micropython?
  - Lacks key embedding features
  - Binary size - micropython 620KB binary vs IoTPy 290KB binary
- IoTPy features
  - Lean memory footprint by leveraging NanoLambda Cloud/Edge for bytecode generation
  - Object-oriented VM implementation & first class embedding support
- Security
  - Python VM provides memory protection and container-like isolation

# Deploying Functions



# Evaluation

Configuration	Latency in <i>ms</i>
<i>NanoLambda</i> Local Caching	6.7 <i>ms</i>
<i>NanoLambda</i> No Local Cache	119.5 <i>ms</i>
<i>NanoLambda</i> No Server Cache	220.4 <i>ms</i>
<i>NanoLambda</i> C Handler	0.3 <i>ms</i>

TABLE I  
AVERAGE LATENCY FOR 100 ITERATIONS OF THE HANDLER FUNCTION

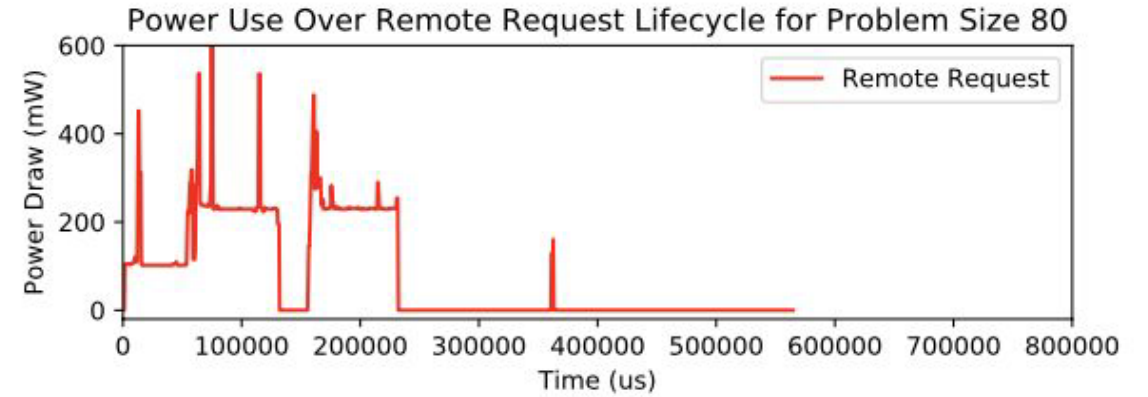
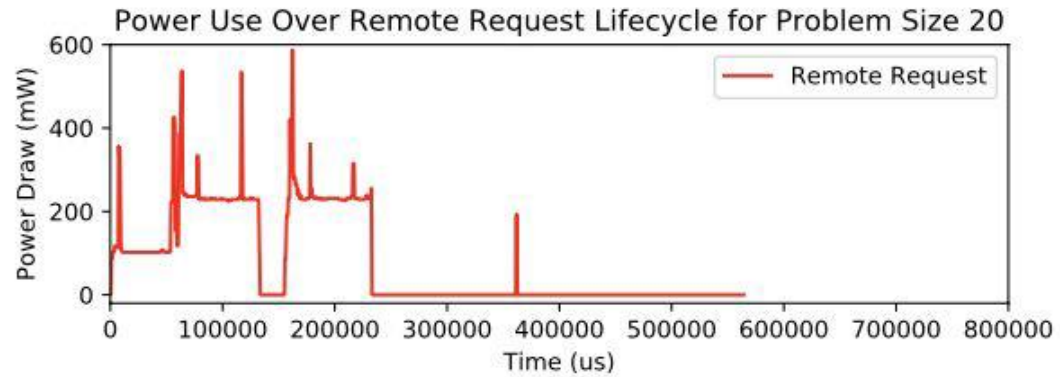
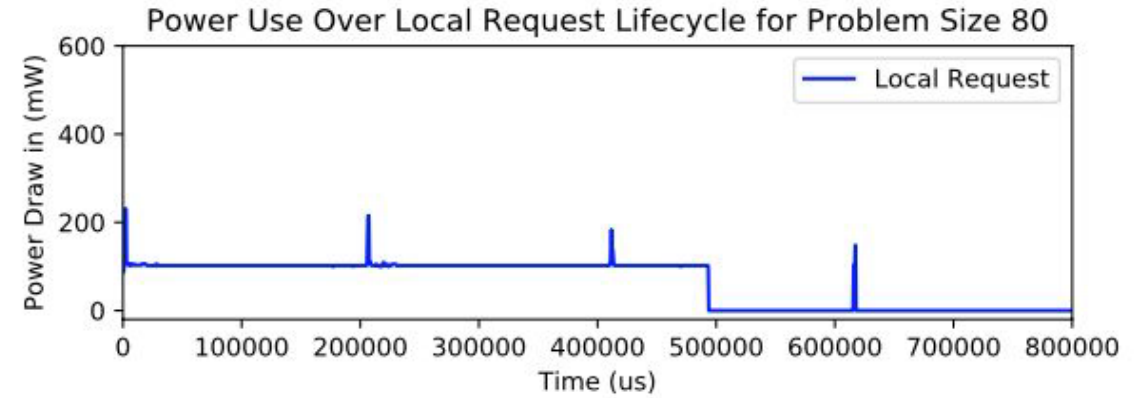
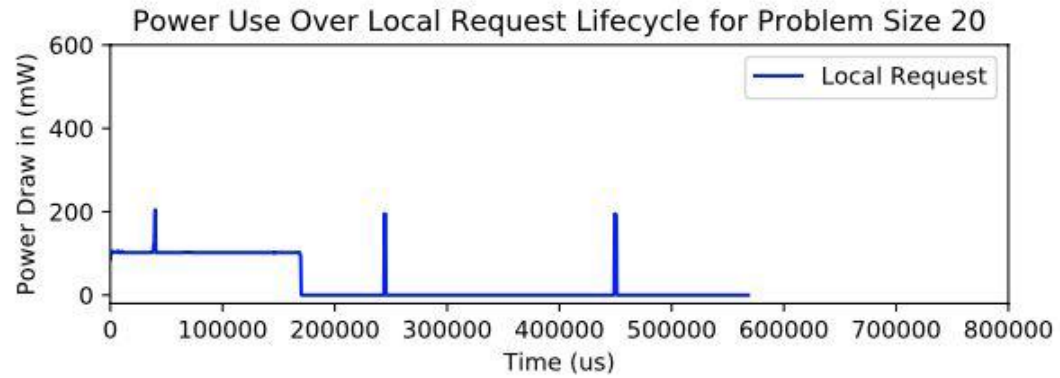
Configuration	Latency ( <i>ms</i> )	Power Use ( <i>mJ</i> )	Memory Use KB
<i>NanoLambda</i> Local Caching	209.4 <i>ms</i>	21.23 <i>mJ</i>	23.6KB
<i>NanoLambda</i> No Local Cache	729.0 <i>ms</i>	85.00 <i>mJ</i>	21.7KB

TABLE II  
AVERAGE TIMES/POWER CONSUMPTION OVER 100 ITERATIONS OF THE HANDLER FUNCTION ON THE CC3220SF MICROCONTROLLER WITH A KS PROBLEM SIZE OF 32 IN EACH CONFIGURATION.

# Predictive Maintenance Application

- Predictive Maintenance is a technique using sensors to detect part failure
- We examine failure detection in motors using accelerometers
- Setup:
  - Accelerometer attached to a motor reads vibration magnitude 5 times a second
  - Data is appended to a Woof for persistence, a history of 32 records is kept.
  - Each append triggers failure detection handler to run
  - Handler is benchmarked running on NanoLambda On Device and NanoLambda Cloud/Edge for various problem sizes and configurations

# Evaluation



# Execution offloading

- NanoLambda On Device is code compatible with NanoLambda Cloud/Edge
  - On Device supports devices as small as ESP8266 and the CC3220SF
  - NanoLambda Cloud/Edge runs on Linux at the edge and in the cloud
- Portability: The choice to use NanoLambda allows for On Device, at the Edge, in the private Cloud, or directly on AWS Lambda

# Naïve scheduler

- Naive algorithm:
  - Pick the lowest latency (time to result) execution strategy based on history
  - Local (On Device) or Remote
- Every 16 invocations reset the history to allow model to recover from network spikes

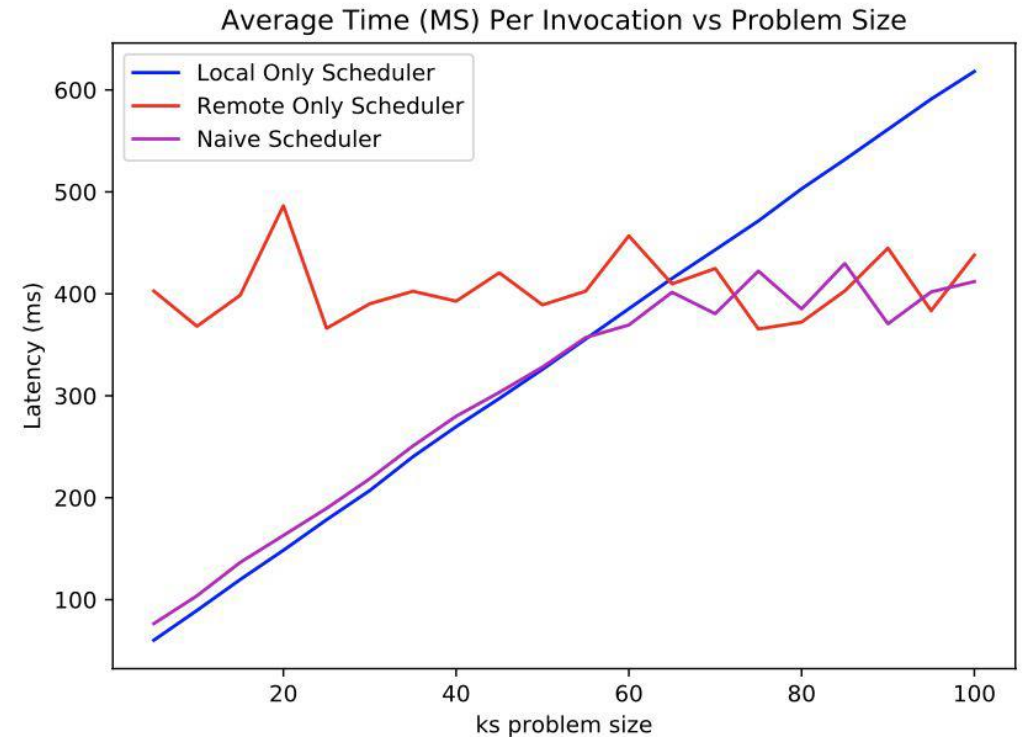


Fig. 6. Comparison of average invocation latency for local invocation strategy, remote invocation strategy, and offloading scheduler invocation strategy.

# Discussion

- Each handler is run in a separate Python VM, common packages may be dynamically linked? – Lightweight Isolation
- Support for Python packages.
- Authentication and authorization