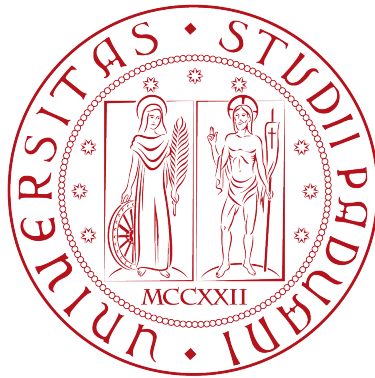


University of Padua

DEPARTMENT OF INFORMATION ENGINEERING  
MASTER DEGREE IN ICT FOR INTERNET AND MULTIMEDIA



Leveraging cloud and machine learning  
technologies for the development of an IOT  
knowledge base

*Master thesis*

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# Summary

This document describes the work done during the 750-hours final project at 221e S.r.l. The project's goal is to architect and develop a cloud-based system capable of ingesting and processing data from heterogeneous IoT sensors so that a knowledge database can be built.

The system must be designed to be scalable and fault-tolerant, and it must be platform-agnostic.

This document is going to describe the company, the idea behind the project, the work done and an assessment of what I developed and learned during my internship.



*“If the past is just dust  
Then the future could be our dream”*

— Lorna Shore

# Acknowledgements

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*Last but not least, I want to give a shoutout to all my friends for having my back and just being there through high and lows. Your friendship means a lot to me, and I appreciate the support and good times we’ve shared.*

*Padova, October 2024*

Alessandro Discalzi



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# Chapter 1

## Introduction

### 1.1 The Company

221e S.r.l.<sup>1</sup> is an Italian company based in Abano Terme (PD). The company, founded in 2012, is a leading supplier of IoT solution providing a wide range of products, both hardware and software, for industrial and wereable applications. The great experience and know how of the company's team, allows to provide clients with the best hardware solution for their needs, which can be enhached via the company's software platform or via a custom solution.



**Figure 1.1:** 221e's logo

### 1.2 Idea

The idea behind the project is to create a new cloud-based software platform for the company's IoT devices. The platform needs to ingest data from a variety of IoT devices, process it and create a knowledge database that can be used to provide insights and predictions to the end user.

### 1.3 Thesis outline

**The second chapter** describes the high level requirements.

**The fourth chapter** describes how the solution has been implemented.

**The fifth chapter** assess the results of the project.

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<sup>1</sup>221e S.r.l. URL: <https://www.221e.com/>.

**The last chapter** describes the conclusion of the project and the future work.

## Chapter 2

# Objectives and requirements

*In this chapter are discussed the main objectives and requirements.*

### 2.1 Cloud Infrastructure

The system must be able to ingest, store and process large amount of IoT data leveraging the power of at least two cloud providers. Furthermore, the final product must be architected to be cloud agnostic, meaning that it must be able to run on any cloud provider. This is important because the system must be able to run on different cloud providers to avoid vendor lock-in and to enable the customer to choose the best cloud provider for their needs.

For the initial development, the system will be developed using Amazon Web Services<sup>1</sup> and Microsoft Azure<sup>2</sup> mainly because of the already developed experience with them, both by the company and by the author of this thesis.

### 2.2 Data collection

The system must be able to ingest data from online devices and on-premise data sources. Online devices are devices that are connected to the internet and can send data to the cloud via MQTT protocol. On-premise data source are offline files that are stored on a local machine and must be uploaded to the cloud. The system must provide a way to upload these files to the cloud.

### 2.3 Data processing

The system must be able to preprocess the data before storage. The preprocessing of the data includes data validation, data cleaning and data transformation, this can be done in cloud or on-premise. The system must be also able to process data after storage. The processing of the data includes data analysis and machine learning operations. The goal of the data processing is to extract useful information from the data and to

---

<sup>1</sup>Amazon Web Services. URL: <https://aws.amazon.com/>.

<sup>2</sup>Microsoft Azure. URL: <https://azure.microsoft.com/>.

provide insights to the customer.

## 2.4 Security

Security is a major concern for the system. In certain scenarios, the data collected could be sensitive and thus must be protected both in transit and at rest.

### 2.4.1 Security in transit

Data in transit are transferred using MQTT protocol which is not encrypted by default. The system must provide a way to encrypt and secure the data in transit. MQTT brokers however supports authentication and authorization through certificates as well as TLS/SSL encryption. Using a broker that supports these features is a must.

### 2.4.2 Security at rest

Data at rest is stored in cloud storage. The cloud storage must provide a way to encrypt the data at rest. The system must also provide a way to manage the encryption keys. Both Amazon Web Services<sup>3</sup> and Microsoft Azure<sup>4</sup> provide a way to encrypt data at rest and manage the encryption keys.

Furthermore, both cloud providers uses a shared responsibility model for security.

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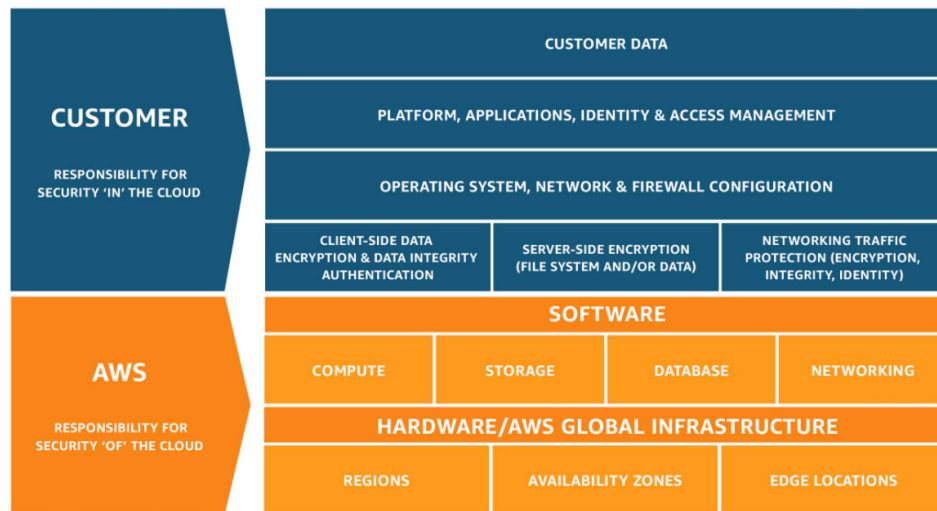
<sup>3</sup>*Amazon Web Services.*

<sup>4</sup>*Microsoft Azure.*

### AWS Shared Responsibility Model

AWS Shared Responsibility Model<sup>5</sup> is a model that defines the responsibilities of AWS and the customer for security. AWS is responsible for the security of the cloud, while the customer is responsible for security in the cloud.

It's important to keep in mind that services in services catalogued as *Infrastructure as a Service* (IaaS) the customer is responsible for the security of the operating system and the applications, while in fully managed services the customer is responsible only for the security of the data.



**Figure 2.1:** AWS Shared Responsibility Model

<sup>5</sup> *AWS Shared Responsibility Model*. URL: <https://aws.amazon.com/compliance/shared-responsibility-model/>.



### Azure Shared Responsibility Model

Azure Shared Responsibility Model<sup>6</sup> is a model that defines the responsibilities of Microsoft and the customer for security. The responsibility varies depending on whether the service is *Software as a Service* (SaaS), *Platform as a Service* (PaaS), *Infrastructure as a Service* (IaaS) or on premise. Regardless of the deployment type, the customer always retains data, endpoints account and access management responsibilities.

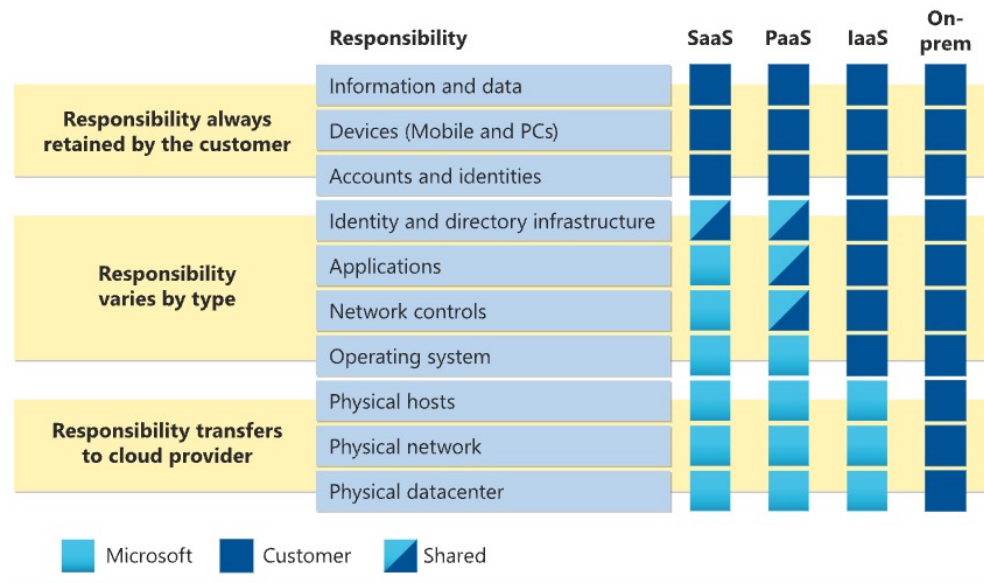


Figure 2.2: Azure Shared Responsibility Model

## 2.5 Scalability

The system must be able to scale horizontally and vertically. The system must be able to handle a large amount of data and must be able to automatically scale to handle more data. Fully managed services and autoscaling properties offered by the cloud providers can be used to handle the scaling of the system.

<sup>6</sup> Azure Shared Responsibility Model. URL: <https://learn.microsoft.com/en-us/azure/security/fundamentals/shared-responsibility>.



## Chapter 3

# Technologies

*In this chapter it's reported the study made on the various technologies taken into account to develop the project.*

### 3.1 Cloud

A deep analysis of the cloud services useful for the project has been made. This section describes the services offered by Amazon Web Services<sup>1</sup> and Microsoft Azure<sup>2</sup>, exploring their features, Pros: and Cons:. Services features are better described in the respective official AWS<sup>3</sup> and Azure<sup>4</sup> documentation, while Pros: and Cons: are based on the author's and the company experience as well as other users reviews that can be found in TrustRadius<sup>5</sup> website.

These services are the most used cloud services in the world and they offer a wide range of services that can be used to develop the project. It's also important to mention that these two providers were chosen right away because of the already developed experience with them, both in the company and in the author of this thesis.

#### 3.1.1 Amazon Web Services

##### Batch

AWS Batch is a fully managed service that enables developers to easily and efficiently run thousands of batch and machine learning computing jobs on AWS.

##### Pros:

- Fully managed
- Scalable
- Cost effective
- Supports different batch processing scenarios

---

<sup>1</sup>Amazon Web Services.

<sup>2</sup>Microsoft Azure.

<sup>3</sup>AWS Documentation. URL: <https://docs.aws.amazon.com/>.

<sup>4</sup>Azure Documentation. URL: <https://docs.microsoft.com/en-us/azure/>.

<sup>5</sup>Trust Radius. URL: <https://trustradius.com/>.

- Supports machine learning
- Easy to use
- Versatile

**Cons:**

- Not well documented

**Bedrock**

AWS Bedrock is a fully managed service that simplifies the deployment and management of machine learning models.

Using AWS Bedrock users can choose from a variety of pre-trained models and deploy them on the edge.

**Pros:**

- Fully managed
- Flexible
- Native support for Retrieval Augmented Generation (RAG) models

**Cons:**

- Costly
- New dependencies may introduce problems
- The chosen model may not be future proof

**DynamoDB**

AWS DynamoDB is a fully managed NoSQL database service that provides fast and predictable performance with seamless scalability. Tables can store and retrieve virtually any amount of data, serving any level of request traffic. It automatically spreads the data and traffic for the table over a sufficient number of servers to handle the request capacity specified by the customer and the amount of data stored, while maintaining Consistent and fast performance.

**Pros:**

- Fully managed
- Fast and predictable performance
- Scalable
- Highly available
- NoSQL

**Cons:**

- Hard to make changes against bulks of records
- Need to know at prior which queries will be made

**Elastic map reduce (EMR)**

AWS EMR is a big data platform that simplifies the deployment and management of big data frameworks, like Apache Hadoop and Apache Spark, on AWS.

**Pros:**

- Fully managed
- Scalable
- Petabyte scale data processing
- Easy resources provisioning
- Reconfigurable

**Cons:**

- Complexity
- Costly

**Glue**

AWS Glue is a fully managed ETL service that enables efficient data integration on a large scale.

**Pros:**

- Fully managed
- Pay per use
- Scalable
- Provides a centralized metadata repository
- Supports different data sources and formats
- Can automatically discover and catalog data from various sources
- Allow for job scheduling
- Data encryption

**Cons:**

- Costly for high workloads
- Performance issues with large datasets
- Complexity

**Greengrass**

AWS Greengrass is an open source edge runtime and cloud service used to build, deploy, and manage device software. It enables the devices to process the data locally, while still using the cloud for management, analytics, and durable storage.

It also enables encryption at rest and in transit and it can also extend device functionality with AWS Lambda functions.

**Pros:**

- Edge computing
- Encryption at rest and in transit
- Extend device functionality with AWS Lambda functions
- ML models deployment

**Cons:**

- Restrained to AWS services
- Not platform agnostic
- Resource intensive for small devices
- Need a connection for the initial setup

**IoT Core**

AWS IoT Core is a fully managed cloud service that lets connected devices easily and securely interact with cloud applications and other devices. It is composed of multiple services like Device Management, Device Defender, Device Advisor, and IoT Analytics and only some of them can be used during the development.

**Pros:**

- Composed of multiple services so only the necessary ones can be used
- Encryption at rest and in transit
- Supports MQTT, HTTP, and WebSockets
- Allows for device management
- Allows for machine learning at edge
- Can trigger events thanks to custom rules

**Cons:**

- Not platform agnostic if installed on devices
- Lacks of integration for some devices

**Kendra**

AWS Kendra is a fully managed enterprise search service that allows developers to add search capabilities across various content repositories leveraging on built in connectors.

**Pros:**

- Fully managed
- Scalable
- Supports multiple data sources
- Easy to use and set up
- Accurate search results

**Cons:**

- Costly

**Kinesis Data Firehose**

AWS Kinesis Data Firehose is a fully managed service that simplifies the process of capturing, transforming and loading streaming data. It acts as an ETL service that can capture, transform, and load streaming data into a variety of AWS services. Additionally it can transform raw data in column oriented data formats like Apache Parquet<sup>6</sup>

**Pros:**

- Fully managed
- Can read data from IoT core and Kinesis Data Streams
- Scalable
- Can transform data
- Can load data into different AWS services
- Supports batching based on time or size

**Cons:**

- Not always cost effective
- Limited transformation capabilities
- Does not support batching based on more complex rules

---

<sup>6</sup>Apache Parquet. URL: <https://parquet.apache.org/>.

**Kinesis Data Streams**

AWS Kinesis Data Stream is a fully managed service that simplify the capture, processing and loading of streaming data in real time at any scale thus enabling real-time data analytics with ease.

**Pros:**

- Fully managed
- Scalable
- Real-time and fast data processing
- Keeps data for 24 hours by default

**Cons:**

- Not always cost effective
- Limited data retention
- Limited data transformation
- Not useful for certain batch processing scenarios

**Lake Formation**

AWS Lake Formation is a fully managed service that simplifies the creation, security and management of data lakes. It allows for cleaning and transforming the data using machine learning.

**Pros:**

- Fully managed
- Scalable
- Secure
- Simplifies lake creation
- Simplifies ingestion management
- Simplifies permission management
- Provides data auditing
- Supports machine learning
- Supports data cataloging

**Cons:**

- Complexity
- Costly
- Not native support for all data sources



### Lambda

AWS Lambda is an event driven serverless compute service that automatically manages the underlying compute resources. AWS Lambda can be used to extend other AWS services with custom logic, and to create new back-end services that can operate at AWS scale, performance, and security.

**Pros:**

- Fully managed
- Serverless
- Pay per use
- Scalable
- Easy to integrate with other AWS services
- Supports multiple programming languages
- Easy to deploy and maintain
- Can run parallel executions
- Low time to market
- Supports custom libraries

**Cons:**

- Limited execution time (15 mins)
- Limited memory
- Limited environment variables
- Maximum 1000 concurrent executions
- Cold start
- Not cost effective for high workloads

### Managed Service for Apache Flink (MSF)

AWS MSF is a fully managed service that simplifies the creation and the execution of real time application using Apache Flink<sup>7</sup>.

**Pros:**

- Fully managed
- Scalable
- Supports batch and stream processing
- Real-time processing
- Large-scale data processing

**Cons:**

- Complexity

---

<sup>7</sup>Apache Flink. URL: <https://flink.apache.org/>.

**Managed Streaming for Kafka (MSK)**

AWS MSK is a fully managed service that simplifies the setup, the scaling and the management of Apache Kafka<sup>8</sup> clusters.

**Pros:**

- Fully managed
- Scalable
- Cost effective
- Secure
- High availability
- Easy to integrate with other AWS services

**Cons:**

- Local testing challenges: hard to replicate the same environment in production and locally
- Not suitable for high traffic scenarios
- Complexity

**Sage Maker**

AWS Sage maker is a cloud-based machine learning platform that allows developer to build, train and deploy machine learning models.

**Pros:**

- Fully managed
- Scalable
- Supports multiple machine learning frameworks
- Supports multiple programming languages
- Allow for easy model deployment

**Cons:**

- Cannot schedule training jobs
- Costly for high workloads

---

<sup>8</sup>Apache Kafka. URL: <https://kafka.apache.org/>.

### Simple Storage Service (S3)

AWS S3 is an object storage service offering scalability, data availability, security, and performance. With S3, any amount of data can be stored and retrieved from anywhere on the web. Data is stored as objects in buckets, with each object representing a file and its metadata.

**Pros:**

- Scalable
- Highly available
- Secure
- Durable
- Cost effective
- No bucket size limit
- No limit to the number of objects that can be stored in a bucket
- Has different storage classes to fit frequent access, infrequent access, and long-term storage

**Cons:**

- Not suitable for small files
- Object size limit (5TB)
- Maximum 100 buckets per account
- Max 5GB per file upload via PUT operation

### 3.1.2 Microsoft Azure

#### Blob Storage

Azure Blob Storage is a fully managed object storage service that is highly scalable and available. It can store large amounts of unstructured data, making it suitable for a wide range of workloads.

**Pros:**

- Fully managed
- Scalable
- Highly available
- Secure
- Cost effective
- No limit to the number of objects that can be stored in a container
- Has different storage tiers to fit frequent access, infrequent access, and long-term storage

- Different storage options (Blob, archive, queue, file and disk)

**Cons:**

- Not suitable for small files
- Object size limit (4TB)
- Maximum 2PB per account in US and Europe
- Maximum 500TB per account in other regions

**Cosmos DB**

Azure Cosmos DB is a fully managed NO-SQL database service supporting multiple data models. It supports multiple NoSQL databases like PostgreSQL, MongoDB, and Cassandra.

**Pros:**

- Fully managed
- Scalable
- Multi model
- Global distribution
- Consistency levels based on the application needs
- Easy to use

**Cons:**

- Expensive
- Slow for complex queries

**DataBricks**

Azure Databricks is a fully managed Apache Spark-based analytics platform supporting a variety of libraries and languages.

**Pros:**

- Fully managed
- Scalable
- Supports multiple programming languages (Python, R, Scala, SQL, Java)
- Supports multiple libraries (Tensorflow, PyTorch, Scikit-learn, etc.)
- Open data lakehouse

**Cons:**

- Costly
- Complexity
- Hard to configure

**Data Explorer**

Azure data explorer is a fully managed, real-time and high volume data analytics service. It offers speed and low latency, being able to get quick insights from raw data.

**Pros:**

- Fully managed
- Scalable
- Real-time data processing
- Low latency
- Supports multiple data sources
- Supports structured, semi-structured and unstructured data
- Fast data ingestion
- Can use batch processing

**Cons:**

- Complexity
- Costly
- Limited capabilities for data transformation
- Hard configuration

**Data Factory**

Azure data factory is a fully managed cloud-based data integration service. It provides tools to orchestrate data workflows while monitoring executions.

**Pros:**

- Fully managed
- Scalable
- Can perform data Analytics using Synapse

**Cons:**

- Complexity
- Limited transformation capabilities

**Data Lake Storage**

Azure Data Lake Storage is a secure and scalable data lake platform. It provides a single place to store structured and unstructured data, making it easy to perform big data analytics.

**Pros:**

- Fully managed

- Scalable
- Secure
- Cost effective
- Compatible with Hadoop
- Supports Python for data analytics

**Cons:**

- Data governance challenges

**Event Hubs**

Azure Event Hubs is a fully managed, real-time data ingestion service that is simple, secure, and scalable. It can be used to stream millions of events per second with low latency, from any source to any destination. It offers also native support for Apache Kafka, allowing user to run existing Kafka applications.

**Pros:**

- Fully managed
- Scalable
- Secure
- Low latency
- Supports Apache Kafka
- Schema registry: centralize repository for schema management
- Real time data processing

**Cons:**

- Costly
- Complexity
- Limitation in event storage
- Consumers need to manage their state of processing

**Functions**

Azure functions is a serverless compute service that enables developers to run code in response to events without the need to manage the infrastructure.

**Pros:**

- Fully managed
- Pay per use
- Scalable
- Supports multiple programming languages

- Easy to deploy and maintain
- Can run parallel executions
- Low time to market
- Supports custom libraries

**Cons:**

- Limited execution time (10 mins)
- Cold start
- Not cost effective for high workloads

**IoT Hub**

Azure IoT Hub is a cloud service that serves as the bridge between IoT devices and solutions in the cloud, facilitating reliable and secure communication. It can handle and manage a large number of devices making it suitable both for small-scale and enterprise-level solutions.

**Pros:**

- Secure
- Supports MQTT, AMQP, and HTTP
- Allows for device management
- Allows for machine learning at edge
- Can trigger events thanks to custom rules
- Can extend device functionality with Azure Functions

**Cons:**

- Not platform agnostic if installed on devices
- Lacks of integration for some devices
- Not well documented
- Costly

**Machine Learning**

Azure Machine Learning is a fully managed service that allows developers to build, train, and deploy machine learning models.

**Pros:**

- Fully managed
- Scalable
- Supports multiple machine learning frameworks
- Supports multiple programming languages

- Allow for easy model deployment
- Cost effective
- Has MLOps capabilities
- Pay as you go

**Cons:**

- Cost raises when training big models
- Hard to optimize

## 3.2 Present Solutions

In this section are presented the solutions that are currently available on the market.

### 3.2.1 Alleantia IoT Edge Hub

IoT Edge Hub is Alleantia<sup>9</sup>'s plug and play solution for the industrial IoT. It offers a wide range of features like device management, alarms and events, log management and report generation. It also supports the integration with Microsoft Azure<sup>10</sup>.

**Pros:**

- Plug and play
- Device management
- Alarms and events
- Log management
- Report generation
- Integration with Microsoft Azure

**Cons:**

- Not platform agnostic
- Does not support Amazon Web Services<sup>11</sup>

### 3.2.2 Eclipse Kura

Eclipse Kura<sup>12</sup> is an open source IoT Edge Framework that serves as a platform for building IoT gateways. It's based on Java/OSGi and it provides API access to the hardware interfaces of IoT Gateways.

**Pros:**

- Open source

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<sup>9</sup> Alleantia. URL: <https://www.alleantia.com/>.

<sup>10</sup> Microsoft Azure.

<sup>11</sup> Amazon Web Services.

<sup>12</sup> Eclipse Kura. URL: <https://eclipse.dev/kura/>.



- Platform agnostic
- Allows for flexible and modular development
- API access to hardware interfaces
- Introduces AI capabilities at the edge

**Cons:**

- Computational complexity for small devices
- Not well documented
- No native support for cloud services

### 3.2.3 Eurotech Everyware Cloud

Eurotech<sup>13</sup> Everyware Cloud is a cloud-based IoT Integration Platform with a microservices architecture that allows to connect, configure and manage IoT gateways and devices.

**Pros:**

- Cloud-based
- Microservices architecture
- Allows to connect, configure and manage IoT gateways and devices
- Supports multiple protocols
- Supports multiple cloud services

**Cons:**

- Last update in 2019

### 3.2.4 STMicroelectronics X-Cube Cloud

STMicroelectronics<sup>14</sup> X-Cube Cloud is a software package that enables the connection of STM32 microcontrollers to the cloud.

**Pros:**

- Supports multiple cloud services
- Offers generic and secure connection to the cloud
- Supports multiple protocols

**Cons:**

- Specific for STM32 microcontrollers
- Specific packages for each cloud service if you want to use the full potential
- The generic solution only runs on a subset of STM32 microcontrollers

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<sup>13</sup>*Eurotech*. URL: <https://www.eurotech.com/>.

<sup>14</sup>*STMicroelectronics*. URL: [https://www.st.com/content/st\\_com/en.html](https://www.st.com/content/st_com/en.html).

### 3.3 Machine Learning at edge

This section describes the technologies that can be used to build and deploy machine learning models at the edge and the Federated Learning approach.

#### 3.3.1 Tensorflow Lite

Tensorflow Lite<sup>15</sup> is the mobile and edge version of Tensorflow<sup>16</sup> that allows to run machine learning models on edge devices.

**Pros:**

- Lightweight
- Supports multiple platforms
- Easy to use
- Well documented

**Cons:**

- Need to pretrain the model before deploying it
- On-device training is limited to Unix-based systems

#### 3.3.2 Tiny Engine

Tiny Engine<sup>17,18</sup> is a machine learning framework that allows to build, train and deploy machine learning models on edge devices.

**Pros:**

- Lightweight
- Supports multiple platforms
- Converts Tensorflow Lite models to C++
- Uses pre trained models

**Cons:**

- Need to pretrain the model before deploying it
- Better support for MCUnet<sup>19</sup>
- Not well documented for deployment of custom models

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<sup>15</sup> *TensorFlow Lite*. URL: <https://www.tensorflow.org/lite>.

<sup>16</sup> *TensorFlow*. URL: <https://www.tensorflow.org/>.

<sup>17</sup> *TinyEngine*. URL: <https://github.com/mit-han-lab/tinyengine>.

<sup>18</sup> Ji Lin et al. “On-Device Training Under 256KB Memory”. In: *arXiv:2206.15472 [cs]* (2022). URL: <https://arxiv.org/abs/2206.15472>.

<sup>19</sup> Ji Lin et al. “Mcunet: Tiny deep learning on iot devices”. In: *Advances in Neural Information Processing Systems* 33 (2020). URL: <https://arxiv.org/abs/2007.10319>.

### 3.3.3 Federated Learning and Transfer Learning

Federated Learning is a machine learning approach that trains an algorithm across multiple decentralized edge devices or servers holding local data samples, without exchanging them. This approach allows for privacy preservation and data security, as well as for the possibility to train models on devices with low computational power as described in “EdgeFed: Optimized Federated Learning Based on Edge Computing”<sup>20</sup>. Another important approach that has been taken into account is Transfer Learning, a technique that transfer the knowledge of a model trained on a specific task to a new task, improving the performance of the model and reducing the time and resources needed to train it, as described in “Federated learning for IoT devices: Enhancing TinyML with on-board training”<sup>21</sup>.

**Pros:**

- Privacy preservation
- Data security
- No need for data centralization
- Low computational power needed
- Low latency

**Cons:**

- Complex to implement
- Limited capabilities

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<sup>20</sup>Yunfan Ye et al. “EdgeFed: Optimized Federated Learning Based on Edge Computing”. In: *IEEE Access* 8 (2020), pp. 209191–209198. doi: 10.1109/ACCESS.2020.3038287.

<sup>21</sup>M. Ficco et al. “Federated learning for IoT devices: Enhancing TinyML with on-board training”. In: *Information Fusion* 104 (2024), p. 102189. issn: 1566-2535. doi: <https://doi.org/10.1016/j.inffus.2023.102189>. URL: <https://www.sciencedirect.com/science/article/pii/S1566253523005055>.



## Chapter 4

# Methodology

### *Introduction*

#### 4.1 Data Collection

#### 4.2 Architecture



## Chapter 5

# Results

*Chapter intro*

### 5.1 Tests





## Chapter 6

# Conclusion

6.1 Objectives achieved

6.2 Future developments

6.3 What I learned

6.4 Final considerations



## Chapter 7

# Bibliography

### Article references

- Ficco, M. et al. “Federated learning for IoT devices: Enhancing TinyML with on-board training”. In: *Information Fusion* 104 (2024), p. 102189. ISSN: 1566-2535. DOI: <https://doi.org/10.1016/j.inffus.2023.102189>. URL: <https://www.sciencedirect.com/science/article/pii/S1566253523005055> (cit. on p. 25).
- Lin, Ji et al. “Mcnnet: Tiny deep learning on iot devices”. In: *Advances in Neural Information Processing Systems* 33 (2020). URL: <https://arxiv.org/abs/2007.10319> (cit. on p. 24).
- Lin, Ji et al. “On-Device Training Under 256KB Memory”. In: *arXiv:2206.15472 [cs]* (2022). URL: <https://arxiv.org/abs/2206.15472> (cit. on p. 24).
- Ye, Yunfan et al. “EdgeFed: Optimized Federated Learning Based on Edge Computing”. In: *IEEE Access* 8 (2020), pp. 209191–209198. DOI: 10.1109/ACCESS.2020.3038287 (cit. on p. 25).

### Website references

- 221e S.r.l.* URL: <https://www.221e.com/> (cit. on p. 1).
- Alleantia.* URL: <https://www.alleantia.com/> (cit. on p. 22).
- Amazon Web Services.* URL: <https://aws.amazon.com/> (cit. on pp. 3, 5, 9, 22).
- Apache Flink.* URL: <https://flink.apache.org/> (cit. on p. 15).
- Apache Kafka.* URL: <https://kafka.apache.org/> (cit. on p. 16).
- Apache Parquet.* URL: <https://parquet.apache.org/> (cit. on p. 13).
- AWS Documentation.* URL: <https://docs.aws.amazon.com/> (cit. on p. 9).
- AWS Shared Responsibility Model.* URL: <https://aws.amazon.com/compliance/shared-responsibility-model/> (cit. on p. 6).
- Azure Documentation.* URL: <https://docs.microsoft.com/en-us/azure/> (cit. on p. 9).

*Azure Shared Responsibility Model*. URL: <https://learn.microsoft.com/en-us/azure/security/fundamentals/shared-responsibility> (cit. on p. 7).

*Eclipse Kura*. URL: <https://eclipse.dev/kura/> (cit. on p. 22).

*Eurotech*. URL: <https://www.eurotech.com/> (cit. on p. 23).

*Microsoft Azure*. URL: <https://azure.microsoft.com/> (cit. on pp. 3, 5, 9, 22).

*STMicroelectronics*. URL: [https://www.st.com/content/st\\_com/en.html](https://www.st.com/content/st_com/en.html) (cit. on p. 23).

*TensorFlow*. URL: <https://www.tensorflow.org/> (cit. on p. 24).

*TensorFlow Lite*. URL: <https://www.tensorflow.org/lite> (cit. on p. 24).

*TinyEngine*. URL: <https://github.com/mit-han-lab/tinyengine> (cit. on p. 24).

*Trust Radius*. URL: <https://trustradius.com/> (cit. on p. 9).