

D4.2 – PROGRESS REPORT ON PLATFORM IMPLEMENTATION AND PETS INTEGRATION

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Executive Summary

Rather than several standalone modules, the PAPAYA project aims at developing an integrated platform for privacy preserving data analytics to make them available in a broad spectrum of products and services, with usable, friendly and accessible safeguards options. The main goal of the platform is to be used by service developers to deploy and run privacy-preserving services, and by service consumers, that are interested to employ privacy-preserving analytics. In the previous deliverable (i.e. D4.1 [1]) we presented a first version of the platform functional design, architecture, and deployment. In particular, (1) we described the design of the main platform components that were elicited based on the requirements presented in D2.2 [2], (2) we explained how different privacy preserving primitives being developed in WP3 (see D3.1 [3] and D3.3 [4]) are integrated into the platform in a way that they will be interoperable/compatible with each other and could work together in the integrated platform; and (3) we presented the design of platform dashboards that provide the UI, configuration functionality, and visualization functionality. In this deliverable we describe changes to the services specified in D4.1 [1] and provide description of several new services being created on top of the approaches developed in WP3 and described in detail in D3.3 [4].

As already mentioned in D4.1 [1], PAPAYA platform services are specified based on the four generic usage scenarios, namely upload model, create model, apply model and collaborative training. In upload model, an already trained machine learning (ML) model can be uploaded to the PAPAYA platform when the client wants to delegate the computationally intensive task (which is applying a model on the client's sensitive data) in a privacy-preserving manner. The create model is used when the client is not able to create the ML model; therefore, the PAPAYA platform generates a model on the protected data shared by the client. Apply model is the use case where already uploaded or created model is applied on the client's protected data in a privacy-preserving manner. Finally, in collaborative training, two or more participants perform a ML training collaboratively while preserving the privacy of the training data.

The PAPAYA platform consists of the following groups of services:

- Privacy-preserving analytics defined in the deliverables D3.1 [3] and D3.3 [5] such as classification on Neural Networks, privacy-preserving clustering, privacy-preserving statistics, and privacy-preserving collaborative training of Neural Networks
- Security and transparency services, including the identity access management (IAM) for authentication and authorization services to the different components that will be integrated in the PAPAYA platform, auditing to support auditing and towards being able to hold stakeholders accountable for their use of PAPAYA, and key manager for managing the cryptographic material during the whole lifecycle of the PAPAYA project in the cases where it will be required.
- The PAPAYA platform provides two dashboards for configuration and visualization, namely the platform dashboard and the agent dashboard. The platform dashboard is used for configuration and monitoring the services provided by the platform whereas the agent



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dashboard is used for viewing the data processing logs from an agent and for showing the configuration of the agent.

 Data subject toolbox services provide a number of mostly independent tools (Explaining Privacy-preserving Analytics, Data Disclosure Visualization, Annotated Log View and Privacy Engine) which provide versatile services related to the data subject privacy. Moreover, the PAPAYA platform provides means to its clients to integrate one or more tools from the provided toolbox to generate an integrated *data subject dashboard*.

As a proof of concept usage, the platform will be deployed on IBM Kubernetes cloud service. In addition to that, all services are designed to be generic in order to be deployed on any other cloud platforms.

The final platform implementation and PETs integration after validation on project use cases, including final version of the dashboard will be given in deliverable D4.3 [6].



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Glossary of Terms

2PC	Two-Party Computation
	Auditing Agent
ABY	Arithmetic sharing, Boolean sharing and Yao's garbled circuits framework
AC	Auditing Collector
	Annotated Log view Tool
	Application Programming Interface
RF	Bloom Filters
BGV	Brakerski-Gentry-Vaikuntanathan homomorphic encryption scheme [7]
BEV	Brakerski/Ean-Vercauteren fully homomorphic encryption scheme [8, 9]
CA	Certificate Authority
CKKS	Cheon-Kim-Kim-Song fully homomorphic encryption scheme [10]
CLI	Command Line Interface
CNN	Convolutional Neural Network
CPU	Central Processing Unit
CR	Container Registry
DB	Data Base
DC	Data Controller
DP	Differential Privacy
DNN	Deep Neural Network
DS	Data Subject
DSRM	Data Subject Rights Manager
DVT	Disclosure Visualization Tool
ECG	Electro cardiogram
ES	ElasticSearch
FC	Fully Connected
FE	Functional Encryption
FHE	Fully Homomorphic Encryption
GRU	Gated Recurrent Unit
HE	Homomorphic Encryption
IAM	Identity Access Manager
KM	Key Manager
ML	Machine Learning
MLP	Multilayer Perceptron
NN	Neural Network
PE	Privacy Engine
PHE	Partially Homomorphic Encryption
PoC	Proof of Concept
PPM	Privacy Preferences Manager
RAM	Random Access Memory



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- REST Representational State Transfer
- ReLU Rectified Linear Unit
- RNN Recurrent Neural Network
- SIMD Single Instruction Multiple Data



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1 Introduction

1.1 Purpose and Scope

The purpose of this deliverable is to provide a comprehensive description of the PAPAYA platform architecture. This deliverable extends the platform design and architecture already presented in D4.1 [1]. In particular, it provides modifications to already described services as well as specification of several new services developed on top of the analytics developed in WP3 and presented in D3.3 [5]. The intended audience for this document consists of two main groups: (1) service developers, who could use the document to understand the design of the existing services and to develop and deploy their own services; (2) service users, who could use the document to understand how to incorporate them in their applications.

It is important to note that the description of the underlying algorithms employed by the services presented in this document is not in the scope of this deliverable (they are described in D3.1 [3] and in D3.3 [5]).

1.2 Structure of the Document

The rest of the document is organized as follows:

- Section 2 provides a high-level overview of the PAPAYA framework, including description of main stakeholders.
- Section 3 describes in detail the core PAPAYA services, which are defined in the deliverable D3.1 [3] and deliverable D3.3 [5]. Provided services in the scope of this deliverable are: (1) Privacy-preserving classification on Neural Networks; (2) Privacy-preserving collaborative training of Neural Networks; (3) Privacy-preserving clustering; and (4) Privacy-preserving statistics.
- Section 4 gives details about how security and transparency is achieved in the platform, including authentication, authorization, auditing and key management for cryptographic tools (if it is required).
- Section 5 presents PAPAYA dashboards, namely platform dashboard and agent dashboard.
- **Section 6** dedicated to Data Subject Toolbox, which provides versatile services (Explaining Privacy-preserving Analytics, Data Disclosure Visualization, Annotated Log View and Privacy Engine) related to the data subject privacy. Tools presented in this section can then be used to form a *data subject dashboard*.
- In Section 7 we describe how to deploy and run all platform components.
- We conclude the deliverable in **Section** 8.



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2 PAPAYA FRAMEWORK

2.1 Main Stakeholders

On a high-level, there are four main stakeholders in the PAPAYA framework:

- 1. *Platform clients:* stakeholders who wish to perform some analytics in a privacy preserving manner. *Platform clients* can be considered as Data Controllers or external queriers who are allowed to request some analytics results while not being the actual owners of the data.
- 2. *Platform administrators:* responsible for platform administration purposes such as resource allocation or monitoring.
- 3. Service providers: the author of the services available on the platform.
- 4. Data Subjects: end-users (e.g. application user) of the Platform clients.

Details about PAPAYA usage scenarios can be found in D4.1 [1].

2.2 PAPAYA framework architecture

The PAPAYA framework (see Figure 1) is composed of two main groups of components: (1) the platform-side components that will be running on the (non-trusted, but semi honest) Kubernetes1 cloud server; and (2) client side components, that will be running on trusted client environment.



Figure 1 PAPAYA Framework architecture

From the functional perspective, the PAPAYA framework components can be regrouped into the following categories (more details about each category can be found in deliverables D4.1 [1] and D3.1 [3]):

¹ <u>https://kubernetes.io/</u>



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- The privacy-preserving analytics services (colored with green in Figure 1) which will allow platform clients to perform analytics of interest, in a privacy-preserving manner. In the first version of PAPAYA we will support the following analytics:
 - 1. Privacy-preserving NN classification. We will provide four services for applying neural network for the purpose of classification in a privacy-preserving manner: (1) 2PC-based; (2) PHE-based; (3) FHE-based; and (4) Hybrid approach. Each one could be preferable than others in different settings, mainly depending on NN architecture
 - 2. Privacy-preserving collaborative training of NN. The service allows multiple participants to perform a ML training collaboratively, while preserving the privacy of the training data.
 - 3. Privacy-preserving trajectory clustering. The service provides means to cluster trajectories in a privacy preserving manner.
 - 4. Privacy-preserving basic statistics. The service provides means for privacy-preserving computation of statistics using functional encryption and privacy-preserving counting using Bloom Filters.

The users of these services are platform clients who can either be Data Controllers or external queriers (in the case for privacy preserving trajectory clustering who are authorized by Data Controllers to request some analytics results. Each service is divided into two parts:

- 1. Server responsible for performing analytics of interest on encrypted data and will run on a PAPAYA's Kubernetes cluster.
- Agent responsible for communication with the appropriate server-side component and responsible for managing cryptographic operations for the client. The agent will be downloaded from the Container Registry (CR) as a Docker image and deployed on a client side. Deployment and execution will be under the responsibility of the platform client.
- The platform security and transparency services (colored with red in Figure 1) which will provide platform authorization, authentication (Identity and Access Management IAM), auditing and cryptographic Key Manager (if needed).
- The PAPAYA dashboards (colored with yellow in Figure 1):
 - 1. **Platform dashboard** will allow: (1) **service providers** to deploy privacy-preserving services; (2) **platform clients** to choose/run privacy preserving services and review operational and auditing logs; and (3) **platform administrators** to configure and manage the platform.
 - 2. **Agent dashboard** will allow **platform clients** to visualize the configuration of the agent running on the client side and review operational and auditing logs.



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• Data Subject Toolbox (colored with blue in Figure 1) consists of a number of mostly independent tools (DS Tool 1: Explaining Privacy-preserving Analytic, DS Tool 2: Data Disclosure Visualization, DS Tool 3: Annotated Log View and DS Tool 4: Privacy Engine), which provide versatile tools for data protection by design by platform clients (acting as data controllers) towards data subjects whose personal data is processed in their services.



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3 Platform Core Services

3.1 Apply Neural Network Model

In this section, we describe several services for applying neural network for the purpose of classification in a privacy-preserving manner. The detailed description of these services can be found in deliverables D3.1 [3] and D3.3 [5].

Prior to using any of the services described in this section, a Neural Network model should be trained locally based on the clear text data with all the required optimization and the dimensionality reduction. After achieving the desired accuracy, the trained model (i.e. architecture and weights) should be saved in a supported format and passed to the service later as described in each of the following subsections.

3.1.1 Privacy-preserving NN classification based on 2PC

In this section, updates on APIs and details on the deployment, configuration and implementation of the privacy preserving NN classification based on 2PC service are presented. Since there is no update on the main components, relationships between these components and the behavioral analysis of this service subsections, they remain the same as introduced in D4.1 [1]. Finally, the evaluation of the service integration is overviewed in Section 3.1.1.4.

3.1.1.1 Deployment and configuration

Deployment. As stated in D4.1 [1], the 2PC-based privacy-preserving NN classification based on two-party computation (2PC) service involves two components, namely the client-side and serverside components as stated in D4.1 [1]. Both of these components are deployed as Docker containers. The server-side component uses a particular NN model for arrhythmia classification. The model is also installed during the deployment of server component. Before its deployment, the server component should receive the information of the local IP address (which should be 0.0.0.0 by default because of the use of Flask framework as a RESTful service), local HTTP port number and the TCP port number (any available port number, except the reserved ports by Ubuntu operating system). Note that the local IP address and the port number is required to run ABY server in the server component.

Configuration. This service does not require any specific configuration for the installed libraries, tools and the operating system that will be used for accomplishing the 2PC-based PP NN classification. On the other hand, once the server-side component is deployed to the cloud environment, it will be assigned a new public IP address and a port number. This newly assigned communication information should be stored in the client-side component to execute 2PC computation. Therefore, the one and only configuration that is required for the client-side component, is accomplished by using "init/" API call.



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3.1.1.2 Implementation constraints

The 2PC-based Privacy-preserving NN classification service uses a slightly modified version of Arithmetic-Boolean-Yao (ABY) as a secure 2PC framework. RESTful Web-service requires Python v3 and Flask v1.0.0 (or higher versions).

3.1.1.3 APIs

Server-side Component APIs:

There is only one RESTful API call at the server-side component in this service and this for initiating the classification in the server side (classify/). The rest of the communication between the server-side component and the client-side component is realized by using ABY sockets.

POST classify/

POST	/classify Classify - Server-side Component	بے
This API ca	II is used for initiating the server-side component for classification.	
Parameters		Try it out
Name	Description	
Number o signals *	nf Number of signals to be classified	
integer (path)	Number of signals - Number of signals to be classified	
Responses		
Code	Description	Links
201	Classification has been initiated in the server-side component	No links
400	invalid input, object invalid	No links



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Client-side Component APIs:

There are two RESTful API calls at the client-side component. The first one is used for initializing the public IP address and the port number of the server (init/) and the second one is used for initiating the classification at the client side. The rest of the communication between the server-side component and the client-side component is realized using ABY sockets.

POST init/

POST /in	it Init - Client-side Component	ب
This API call is us	sed for initializing the port number and the IP address for clien	t-side component.
Parameters		Try it out
Name	Description	
IP address * ^{requi}	IP address of the server	
string (path)	IP address - IP address of the server	
Port number * ^{requir}	port number of the server	
string (path)	Port number - port number of the server	
Responses		
Code Des	cription	Links
201 Ser	ver's IP address and the port number has been recorded.	No links
400 inva	lid input, object invalid	No links



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POST classify/

POST	/classify Classify - Client-side Component	به
This API ca	Il is used for uploading the file to be classified.	
Parameters	3	Try it out
Name	Description	
Input file * requi string (path)	Input file consist ECG signals to be processed. Choose File No file chosen	
Responses		
Code	Description	Links
201	File has been uploaded for classification.	No links
400	invalid input, object invalid	No links

3.1.1.4 Service integration evaluation

We have implemented each component as a Docker container. The client-side component is deployed on the local machine and the server-side component is deployed on the IBM cloud. After the server-side component is deployed to IBM's cloud, the server-side component's public IP address and the port number are stored in the client-side component via the init/ API call. Later, the file to be processed is uploaded to the client-side component using classify/ API call of the client. Then, the client-side component sends the classification request to the server-side component. Once two components are ready for classification, both parties execute classification



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using the ABY framework. We have tested our integration by following the steps illustrated in Figure 2 and the modules were executed correctly and output the expected classification result.



Figure 2 Service Integration Evaluation for PP NN Classification based on 2PC

3.1.2 Privacy-preserving NN classification based on PHE

In this section, updates on APIs and details for the deployment, configuration and implementation of the PHE-based privacy-preserving NN classification-based service are presented. Since there is no update on the main components, relationships between these components and the behavioral analysis of this service subsections, these sections remain the same as introduced in D4.1 [1]. Finally, the evaluation of the service integration for a particular NN model is overviewed in the service integration evaluation Section 3.1.2.4.

3.1.2.1 Deployment and configuration

Deployment. As stated in D4.1 [1], the PHE-based privacy-preserving NN classification service involves two components: the client-side and server-side components. Both of these components are deployed as docker containers. The current version of this service works on particular NN models (ECG classification and MNIST dataset classification). The model is also installed during the deployment of the server-side component. Before its deployment, both the server and the client components may need the information of the local IP address, local HTTP port and local



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TCP port. Note that the local IP address and the local TCP port number are required to ensure communication between these two parties achieved through socket programming.

Configuration. This service does not require any specific configuration for the installed libraries, tools and the operating system that will be used for accomplishing the PHE-based PP NN classification. On the other hand, once the server-side component is deployed to a cloud environment, it will have a new public IP address and a port number. This newly assigned communication information should be also stored in the client-side component to execute classification. Therefore, the one and only configuration is required for the client-side component, which is realized by using "init/" API call.

3.1.2.2 Implementation constraints

The PHE-based privacy-preserving NN classification service uses the Paillier library. Socket programing is used to ensure the communication between the two parties. The REST-full webservice requires Python v3 and Flask v1.0.0 (or higher versions).

3.1.2.3 APIs

Server-side component APIs:

There is only one REST-full API call for the server-side component in this service, namely "classify/". This API call is used for initiating the classification. Once the classification is initiated, then all necessary communications for executing classification are accomplished through socket calls between the server-side and client-side components.



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POST classify/

POST	/classify Classify - Server-side Component	ب
This API ca	Il is used for initiating the server-side component for clas	sification
Parameters		Try it out
Name	Description	
NN mode selection string (path)	 ID of the predefined NN model NN model selection - ID of the predefined NN model 	I
Responses		
Code	Description	Links
201	Server has been initiated for [name] NN Model.	No links
400	invalid input, object invalid	No links



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Client-side component APIs:

There are two REST-full API calls at the client-side component. The first one is used for initializing the public IP address and the port number of the server (init/) and the second one is used to initiate the classification at the client side. Socket calls are used for communication purposes between the server-side and client-side components

POST classify/

POST /classi	fy Classify - Client-side Component	ب
This API call is used fo	r sending the NN model choice and the input file for the c	lassification.
Parameters		Try it out
Name	Description	
NN model selection * ^{required}	ID of the predefined NN model	
string (path)	NN model selection - ID of the predefined NN model	
Input * ^{required}	ID of the predefined NN model	
(path)	Choose File No file chosen	
Responses		
Code Descriptio	n	Links
201 Client has	been initiated for classification.	No links
400 invalid inpu	ıt, object invalid	No links



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POST init/

POST /i	.nit Init - Client-side Component	ب		
This API call is used for initializing the port number and the IP address for client-side component.				
Parameters		Try it out		
Name	Description			
IP address * ^{rec}	IP address of the server			
string (path)	IP address - IP address of the server			
Port number * ^{req}	uired port number of the server			
string (path)	Port number - port number of the server			
Responses				
Code	Description	Linke		
201		No linko		
201 S	erver's IP address and the port number has been recorded.	NO IINKS		
400 ir	nvalid input, object invalid	No links		



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3.1.2.4 Service integration evaluation

Service integration evaluation will be presented in D4.3 [6].

3.1.3 Solution based on Homomorphic Encryption

This solution uses Homomorphic Encryption (HE) in order to build a privacy-preserving neural network inference solution. This solution uses the CKKS scheme [10] implemented in Microsoft, open source, SEAL library [11]. The solution takes a pre-trained neural network composed of the supported layers (Dense, Convolution), and activation functions and provide an encrypted version for secure inference. The detailed design of this service is presented in D4.1 [1].

3.1.3.1 Service integration evaluation

Service integration evaluation will be presented in D4.3 [6].

3.1.4 Privacy-preserving NN classification based on hybrid approach

The hybrid solution uses both, the HE and 2PC, in order to build a privacy preserving NN classification framework and maximize the efficiency of classification on deep NN. The solution is *practically* generic, namely, it supports different types of DNN (i.e., MLP, CNN, and RNN) with any number of layers, any number of neurons in each layer and any activation function (from the set of supported activation functions), while the performance still acceptable (grows linearly with the DNN's depth).

We presented a detailed design in D4.1 [1]. In the following subsection we describe how we integrated and evaluated the service.

3.1.4.1 Service integration evaluation

We evaluated the service integration as depicted in Figure 3. In particular, we developed: (1) an application (ClientApp) which simulates client's application (in our case it reads ECG samples from a file and classify them using the PAPAYA service); (2) the agent (provides basic cryptographic functionality and service interfaces); and (3) the service (which performs the privacy preserving classification). Each of this component is implemented as a docker container. ClientApp and the agent are deployed on a local machine, while the service is deployed on IBM cloud. More details about the API of each component could be found in D4.1 [1]. We run the classification flow for all the inputs and everything works as expected.



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Figure 3 Service integration evaluation



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3.2 Collaborative Training of Neural Network

Collaborative training of NN allows multiple participants to perform a ML training collaboratively, while preserving the privacy of the training data. In the previous version of this document (D4.1 [1]) we presented a complete specification of the service based on Shokri and Shmatikov approach [12]. Since that, in a course of WP3, we have implemented an additional approach for privacy preserving training of DNN based on the method presented by Abadi et. al. [13] (presented in D3.3 [5]). We integrated this solution into the service as well. There are minor changes in almost all previously presented (in D4.1 [1]) components and APIs. So, we provide a complete specification of the service in this section.

- 3.2.1 Main components and their relationships
- 1. The client side responsible for performing the following functionalities:
 - a. NN training
 - b. Adding Differential Privacy (DP) noise
 - i. Train local DP model
 - ii. Noise updates sent to the server
 - c. Uploading noised gradients or DP weights
 - d. Downloading the model parameters to/from the centralized server.



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Figure 4: The client agent components

Figure 4 presents the client agent components that will run on the client side (trusted environment).

- 2. The centralized server (server side) will provide the following functionality:
 - a. Allow participants to define and download the initial model.
 - b. Aggregate the gradients from the collaborative training participants.
 - i. Manage updates statistics and decay the less relevant updates
 - ii. Updates averaging



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c. Allow participants to download the model parameters during the training phase.



Figure 5: The server components

Figure 5 presents the server-side components that will run on PAPAYA K8s cluster. The service instance will be allocated per training task.



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3.2.2 Behavioral analysis:



Figure 6: Collaborative training - sequence diagram

As shown in Figure 6, the training phase begins when the required minimal number of participants (as defined in the configuration) requested to join and start the training. Each client will provide a path to the local dataset to the agent. The agent will perform a NN training locally and upload the noised gradients or weights of noised model (depending on the chosen approach and configuration) to the centralized server (located on the PAPAYA cluster). The server will aggregate the updates to the centralized model. The aggregation techniques can be configured. The server allows agents to download the updated model. Client agents download the model and overwrite the local one. The agent proceeds to the training phase on the updated model and so on. The client agent communicates with the server via REST-API calls. The client app should communicate with the client agent via REST-API calls.



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3.2.3 Deployment and configuration

The necessary information regarding the deployment and configuration of this service is described in D4.1 [1].

3.2.4 Implementation constraints

The service (both the client-side and server-side components) is implemented as a REST-full web service using python² version 3 and Flask framework. Both client and server document the APIs via Swagger³.

The server stores the model in a mounted volume within the K8s server.

As mentioned above, the server side will allocate legal tokens per training task and will expect to receive this token on every request.

The client-side agent expects to receive compiled Keras model.

If the client app chooses to use a differential private local training (based on M. Abadi) approach, then the service supports only the following optimizers:

- sgd
- adam
- adagrad

In addition, in this approach, the client app should provide the model's compilation configuration:

- loss
- metrics
- optimizer
- learning rate

These configurations are used in order to recompile the model to privacy preserving model (using tensorflow/privacy library).

² <u>https://github.com/encryptogroup/ABY</u> - Release: Oct 19, 2019

³ <u>https://github.com/encryptogroup/ABY</u> - Release: Oct 19, 2019



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3.2.5 APIs

Server-side APIs:

Servers http://0.0.0:5000/v1 ~			
Server Privacy preserving collaborative training server APIs	\checkmark		
GET /download_model download the most recent model			
GET /download_weights download the most frequently updated model's weights			
POST /init the API to initiat and to join the collaborative training			
GET /join join collaborative training			
POST /reset server reset			
GET /train join to the training process			
POST /upload_gradients update the centralized model with participant's gradients			

Client agent APIs: Servers

agent Colla	borative training client side agent APIs
GET /ge	et_config retrieve dp and anomaly detection configuration
GET /g	et_model download a NN model
GET /ge	et_status get training status
POST /in	nit initialize regestration with the server.
GET /j	oin join to the collaborative training
POST /se	et_config upload configuration of differential privacy and anomaly detection
POST /se	et_model load NN model and relevant for collaborative training information
POST /se	et_token set authentication token
POST /t:	rain strat a NN training



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3.2.6 Service integration evaluation

We executed the complete deployment and training flow and evaluated the entire privacy preserving training flow on the Swell dataset⁴ (Kaggle, based on use case 2). In particular, we (1) uploaded both, the server-side and the client-side containers to the K8s container registry on IBM cloud; (2) deployed the server-side container on IBM cloud using Papaya platform dashboard; (3) deployed client-side agent and an application which simulates client's application on local machine; and finally (4) executed the entire flow, as depicted in Figure 6. The evaluation results with respect to accuracy/privacy are presented in D3.3 [5].

3.3 Clustering

3.3.1 Privacy-preserving clustering based on 2PC

In D4.1 [1], Privacy-preserving trajectory clustering based on 2PC was briefly introduced. There are no further updates for the main components, relationships between main components and the behavioral analysis subsections in this version. This document consists of updates on deployment, configuration, implementation constraints and APIs sections. Service integration evaluation will be presented in D4.3 [6].

3.3.1.1 Deployment and configuration

Deployment. As stated D4.1 [1], the privacy-preserving trajectory clustering based on two-party computation (2PC) service and consists of two components, namely the client-side and server-side components. Both components are deployed as Docker containers. Before the deployment, the server component should have the information of the local IP address (which should be 0.0.0.0 by default because of the use of Flask framework as a RESTful service), local HTTP port and the local TCP port (any available port numbers, except the reserved ports by Ubuntu operating system). Note that the local IP address and the TCP port number is required to run ABY server in the server component.

Configuration. This service does not require any specific configuration for the installed, libraries, tools and the operating system that will be used for accomplishing the 2PC-based PP clustering. On the other hand, once the server-side component is deployed to a cloud environment, it will have a new public IP address and a port number. This newly assigned communication information should be also stored in the client-side component to execute 2PC computation. Therefore, the one and only configuration is required for the client-side component, which is realized by using "init/" API call.

⁴ <u>https://github.com/encryptogroup/ABY</u> - Release: Oct 19, 2019



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3.3.1.2 Implementation constraints

The privacy-preserving trajectory clustering based on 2PC service uses the latest version of the ABY framework as a 2PC library. For RESTful Web-service, Python v3 and Flask v1.0.0 (or higher versions) are used.

3.3.1.3 APIs

Server-side component APIs:

There is only one RESTful API in the server-side component. This API is used for initiating the ABY based server in the server-side component. All other communications between the client-side component and the server-side component are realized by using ABY sockets.

POST cluster/

POST	/cluster Cluster - Server-side Component	ب	
This API ca	all is used for initiating the server-side component for clustering.		
Parameter	s	Try it out	
Name	Description		
Request	 required Request for initiating the clustering 		
(path)	Request - Request for initiating the clustering		
Responses			
Code	Description	Links	
201		No links	
201	Clustering has been initiated in the server-side component	NO III/KS	
400	invalid input, object invalid	No links	



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Client-side component APIs:

There are two RESTful API calls in the client-side component. The first one is used for initializing the public IP address and the port number of the server (init/) and the second one is used for initiating the clustering in the client side. While initiating the clustering, the to-be clustered line segments obtained from the trajectories are uploaded.

POST init/

POST /init	Init - Client-side Component	ب	
This API call is used for initializing the port number and the IP address for client-side component.			
Parameters	Try i	it out	
Name	Description		
IP address * ^{required}	IP address of the server		
string (path)	IP address - IP address of the server		
Port number * ^{required}	port number of the server		
string (path)	Port number - port number of the server		
Responses			
Code Descript	ion	Links	
201 Server's	IP address and the port number has been saved	No links	
400 invalid in	put, object invalid	No links	



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POST cluster/

POST /cluster Cluster - Client-side Component			
This API ca	Il is used for initiating the client-side component for clustering.		
Parameters	3	Try it out	
Name	Description		
Line segments string (path)	File that consists of the line segments to be clustered. Choose File No file chosen		
Responses			
Code	Description	Links	
201	Clustering has been initiated in the client-side component	No links	
400	invalid input, object invalid	No links	

3.3.1.4 Service integration evaluation

Service integration evaluation will be presented in D4.3 [6].

3.3.2 Privacy-preserving clustering based on MinHash

In this section we provide the design of privacy preserving clustering based on 2PC described in D3.3, based on the MinHash clustering algorithm [14] [15]. The solution implemented uses Facebook Crypten open source library.



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3.3.2.1 Main components and their relationships

In this scenario there are three actors. First, the client that want information about people travels. Second, the Orange Business Service (OBS) that host the service' server. Finally, the Orange Mobile Network (OMN) that provides the requested data to the service.

This solution has two main components:

- 1. Server-side component
- 2. Client-side component

Server-side component:

It consists in the following two modules (Figure 7):

- Agent module. This module is used to communicate with the client agent and the antenna. Also, the module supervises the computation by invoking the cryptographic module.
- Crypto module. The crypto module provides the 2PC framework for the computations.



Figure 7 Server-side component

Client-side component:

It consists in the following modules (Figure 8Error! Reference source not found.):

• Agent module. The agent module deals with the communications with the server agent and OMN. It also supervises the computations by invoking the cryptographic module.


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• Crypto module. This module provides the 2PC framework that will be used during the computations of the clustering.



Figure 8 Client-side component

3.3.2.2 Behavioral Analysis.

The service operates as Figure 9 shows. The client starts by sending a clustering query to the server (step 1). The server analyses the query and then make the relevant request to OMN that gather the necessary data (steps 2, 3). The server agent computes the MinHash of the received data (step 4). Then the server agent encrypts the hashes and proceeds with the computations of the clusters with 2PC (steps 6, 7). Once the clusters are computed, a check for k-anonymity is done (steps 8, 9). Finally, the client decrypts and recovers the result (steps 10, 11).



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Figure 9 Trajectory clustering analysis

3.3.2.3 Deployment and configuration

The components described above will be deployed within the platform as Docker containers.

Configuration parameters of the server include: the IP addresses and port numbers of the two types of clients. In turn, the client should have the server's IP address and port number. Requestor and client do not interact with each other.

3.3.2.4 Implementation constraints

The agents are implemented in Python programming language and communicate using a REST API. The 2PC computations are done using the Facebook's CrypTen⁵ open source library.

⁵ <u>https://github.com/facebookresearch/CrypTen</u>



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3.3.2.5 APIs

<u>Server API:</u> There are two REST API calls supported by the Server-side component: (1) *query;* and (2) request_data. The query call receives the query file from the client and processes it. The request_data call requests data from the providers. Following is a detailed description of the calls.

POST	T /request_data Request the data to the provider.			
Request t	the data to the provider.			
Paramete	ters	Try it out		
Name	Description			
request	Request for the data to the provider in json file.			
0	request - Request for the data to the provider in json file.			



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GET /qu	ery Get the query from the client.			
Get the query fro	Get the query from the cleint and process it.			
Parameters		Try it out		
Name	Description			
query * required (query)	Query from the client in json file. query - Query from the client in json file.			
Responses		Response content type application/json v		
Code Des	cription			
405 Inva	lid input			

<u>Client API:</u> There is only one REST API call on the client side. The call is *query*, it is used to generate and send a query file to the server.



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POST /g	query Send a query for clustering to the server.			
Send a query f	Send a query for clustering to the server.			
Parameters		Try it out		
Name	Description			
query * require	ed Query to the server for clustering.			
(query)	query - Query to the server for clustering.			
Responses		Response content type application/json ~		
Code D	escription			
405 In	avalid input			

3.3.2.6 Service integration evaluation

Service integration evaluation will be presented in D4.3 [6].

3.4 Basic Statistics

3.4.1 Privacy-preserving statistics based on Functional Encryption

This solution uses functional encryption to allow a requestor to compute statistics on users' mobile usage. Functional encryption ensures data confidentiality and also that only the computation agreed by the user can be done. A detailed description of the design is given in D4.1 [1].

3.4.1.1 Service integration evaluation

Service integration evaluation will be presented in D4.3 [6].

3.4.2 Privacy-preserving Counting using Bloom Filters

In this section, we provide the design of the solution for privacy preserving counting using Bloom filters as described in D3.3 [4]. The implementation of this solution uses Microsoft SEAL for the



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homomorphic computations and Facebook Crypten for the 2PC computations: both are open source libraries. The main update from D4.1 [1] is the introduction of 2PC for the computation of the k-anonymity.

3.4.2.1 Main components and their relationships

We give a brief overview of the components to understand the update in the next sections. Full details are available in D4.1 [1].

The solution consists of three main components:

- 1. Requestor-side component. This component is composed of two modules:
 - a. Agent module. The purpose of the agent module is to communicate with server component.
 - b. Crypto module. The goal of the crypto module is to provide cryptographic primitives for the requestor.
- 2. Server-side component. This component is composed of three modules:
 - a. Storing module. The purpose of the storing module is to store on the server the requests sent by the requestor.
 - b. Operation module. This module compute bloom filters intersections/unions in the encrypted domain, depending on the request from the requestor.
 - c. Interface. The goal of the interface module is to deal with communications from the requestor and the client.
- 3. Client-side component. This component is composed of two modules:
 - a. Agent module. The agent module interfaces with the server, collects data from antennas and invokes the crypto module to encrypt the data.
 - b. Crypto module. The crypto module provides the cryptographic primitives for the client.

3.4.2.2 Behavioral analysis

We only give an overview of the behavioral analysis to be able to explain the update. The full details of the behavioral analysis are available in D4.1 [1].

The service operates in three main phases (see Figure 10):

- 1. System initialization, including the analytics initialization and the key generation;
- 2. Encryption phase used to encrypt the Bloom Filters;
- 3. Statistics phase, including the computation on encrypted data and the decryption of the result.

The main update in the new version comes during the Statistics phase. In the previous version, there was no k-anonymity check before the requestor decryption. Now, before sending the result to the requestor for decryption, we apply a mask (Step 11). Then the ciphertext is sent to the



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requestor (Step 13). The requestor decrypts it (Step 14). After that, the server and the requestor check if the value if greater than the anonymity threshold (Step 15, 16), using 2PC. Finally, the requestor decrypts the final result (Step 17).



Figure 10 Privacy-preserving statistics with Bloom Filters

3.4.2.3 Deployment and configuration

The above-described components will be deployed within the platform as Docker containers.

Configuration parameters of the server include: the IP addresses and port numbers of the two types of clients, and in turn the clients should have the server's IP address and port number. Requestor and client do not interact with each other.

3.4.2.4 Implementation constraints

During the collection of encrypted Bloom filters provided by the client, the server may wait for the reception of several Bloom Filters (at different times) before applying the statistics operation.

The input Bloom filters must have all the same size so that operations such as intersection could be defined.



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3.4.2.5 APIs

The APIs are the same, so we refer to D4.1 [1] for the description of this section.

3.4.2.6 Service integration evaluation

Service integration evaluation will be presented in D4.3 [6].

4 Platform Security and Transparency

4.1 IAM

The Identity Access Manager (IAM) carries out the Authentication and Authorization services to protect the accesses to the PAPAYA platform. In order to do so, a traditional bastion-host pattern has been applied, which implies the implementation and deployment of different components, not only for the IAM server itself. Hence this chapter is devoted to describing the IAM component but also those other complementary components necessary to provide these services. In contrast with the previous version of this deliverable (D4.1), the following sections of this report focus on the deployment details of the different components necessary for providing Authentication and Authorization services within the Papaya platform.

4.1.1 Main components and their relationships

The main IAM components and their relationships with other components of the PAPAYA platform is described in detail within deliverable D4.1. Just as a reminder on what is described in the previous version of this document, as shown in the Figure 11, all access to the PAPAYA framework will be done towards the Security Proxy component, which will contact the IAM to verify if every access is authenticated and authorized and, if the access is granted, it will be redirected to the corresponding Computation Component.



Figure 11 IAM and Security Proxy in the PAPAYA Platform



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4.1.2 Deployment and configuration

As mentioned before, this section is devoted to describing the deployment of a set of different components, not just the IAM server itself. In order to provide the Authentication and Authorization services for the PAPAYA framework it is necessary to follow each of these steps:

- Step 1. IAM deployment (including the DDBB associated)
- Step 2. OAuth 2 client configuration
- Step 3. Computation Components deployment
- Step 4. Security-Proxy deployment

The Figure 12 shows these steps in place:





4.1.2.1 Step 1 IAM deployment

As the IAM selected is based on a broadly used open source solution, the deployment of the IAM server can be performed easily using the following command:

```
      $helm install --name keycloak -f keycloak-values.yaml codecentric/keycloak

      NAME: calico-macaw

      LAST DEPLOYED: Thu Mar24 10:32:36 2020

      NAMESPACE: default

      STATUS: DEPLOYED

      RESOURCES:

      ==> v1/ConfigMap

      NAME
      DATA AGE

      calico-macaw-keycloa-sh
      1

      Os

      calico-macaw-keycloa-test
      1

      Os
```



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==> v1/Pod(related) NAME **READY STATUS RESTARTS AGE** calico-macaw-keycloa-0 0/1 ContainerCreating 0 0s ==> v1/Secret NAME TYPE DATA AGE calico-macaw-keycloa-db Opaque 1 Os calico-macaw-keycloa-http Opaque 1 Os ==> v1/Service NAME TYPE CLUSTER-IP EXTERNAL-IP PORT(S) AGE calico-macaw-keycloa-headless ClusterIP None <none> 80/TCP,8443/TCP 0s calico-macaw-keycloa-http ClusterIP 10.98.100.117 <none> 80/TCP,8443/TCP 0s ==> v1/StatefulSet NAME READY AGE calico-macaw-keycloa 0/1 Os NOTES: Keycloak can be accessed: * Within your cluster, at the following DNS name at port 80: calico-macaw-keycloa-http.default.svc.cluster.local * From outside the cluster, run these commands in the same shell: export POD NAME=\$(kubectl get pods --namespace default -l app.kubernetes.io/instance=calico-macaw -o jsonpath="{.items[0].metadata.name}") echo "Visit http://127.0.0.1:8080 to use Keycloak" kubectl port-forward --namespace default \$POD NAME 8080 Login with the following credentials: Username: keycloak To retrieve the initial user password run: kubectl get secret --namespace default calico-macaw-keycloa-http -o jsonpath="{.data.password}" | base64 --decode; echo

As result of the execution of the previous command, the whole information or the containers created and how to access to the IAM is displayed.

4.1.2.2 Step 2. Configure the client

The access from the client side to the platform will be done following an OAuth 2 standard as a client. Therefore, it is necessary to configure the access of the client on the IAM platform beforehand. Although in the PAPAYA deployment, this configuration will be automatized using the appropriate scripts and the client's configuration details can be consulted in the following link: https://www.keycloak.org/docs/9.0/server_admin/#_clients



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4.1.2.3 Step 3. Computation Components Deployment

Once the IAM server is in place and the client has been configured properly it is then the turn of the deployment of all the PAPAYA platform services. Each service deployed can be associated with one or several entry points that need to be exposed to the exterior of the platform. Although how Computation Components are deployed within the platform is not the focus of this section (each Computation Components will have their own deployment plan), the entry points associated with them will be important. This list of entry points will be used in the following configuration steps.

4.1.2.4 Step 4. Security-Proxy deployment

At the last step, using the list of entry points obtained previously, the Security-Proxy can be configured and deployed. In order to do so, it is only necessary to execute the following command:

\$helm install --name security-proxy -f security-proxy-values.yaml .

4.1.3 Implementation constraints

All issues and constraints associated with this set of components are described in the previous version of this document D4.1.

4.1.4 APIs

The API definition of all the services described in this section can be consulted in the following link: <u>https://www.keycloak.org/docs-api/9.0/rest-api/index.html</u>

4.1.5 Service integration evaluation

The deployment of the whole set of components for providing the Authentication and Authorization services has been deployed successfully locally but also within the cloud environment provided for the deployment of the project. Service integration evaluation will be presented in D4.3 [6].

4.2 Auditing

Towards being able to hold stakeholders accountable for their use of the PAPAYA platform and services, data processing is logged both as part of the platform and locally at agents.

4.2.1 Platform auditing

Our approach to platform auditing is described in Section 5.2.1 of D4.1 [1]. Here, we first provide an overview and then describe how to evaluate the integration of the platform auditing into the PAPAYA platform.



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4.2.1.1 Overview

We plan to deploy platform auditing in two stages:

- 1. In the first stage (due M30, as part of the intermediate version of the platform milestone), we deploy the commonly used Elastic stack in the platform's K8s cluster.
- 2. In the second stage due M36, we improve the security of the setup by making the resulting logs (i) tamper proof and (ii) verifiable—in terms of authenticity and time—by third parties.

As part of the Elastic stack, we will in the first stage use the Filebeat⁶, Elasticsearch⁷, and Kibana⁸ components. Containers that run the analytics services (Section 3) log their operations to standard output, and Filebeat is configured in K8s (as part of pods) to collect all that output and send it to the Elasticsearch instances we run as part of the platform for storage. From the platform dashboard, administrators and platform users will be able to access an instance of Kibana to view their logs stored in Elasticsearch. One significant open question is how to deal with access control to logs in Kibana⁹. Our goal is to have this operational as part of the intermediate version of the platform in M30.

The second stage will involve one additional component of the Elastic stack: Logstash¹⁰. Logstash is a log pre-processor that receives logs from Filebeat, performs some processing, and then forwards them to Elasticsearch. We will create a custom Logstash component based on an existing secure logging scheme built for the Elastic stack [16]. The component will efficiently sign the logs such that they can be shared and verified by third parties. The figure below summarizes the flow of logging data for platform auditing and the involved components.



⁶ https://www.elastic.co/beats/filebeat

⁷ https://www.elastic.co/elasticsearch/

⁸ https://www.elastic.co/kibana

⁹ This used to be a paid feature of the Elastic stack but rudimentary functionality is now part of the open source components, see <u>https://www.elastic.co/blog/security-for-elasticsearch-is-now-free</u>. Before M30, we need to decide on using the limited free functionality of paying for the full functionality.

¹⁰ <u>https://www.elastic.co/logstash</u>



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4.2.1.2 Integration evaluation

We will validate the integration of platform auditing as follows for the first stage:

- 1. Create a new user for the platform.
- 2. Create one or more instances of analytic services.
- 3. For at least one service, run its agent and perform some analytics.
- 4. As the newly created user, access Kibana through the platform dashboard. Verify that all expected logs are available.
- 5. Perform the same verification as in 4, but for another user, verifying that it cannot see the newly generated logs.
- 6. Perform the same verification as in 4, but for the platform administrator, ensuring that it can see all logs.

For the second stage, there are two additional steps:

- 7. Repeat steps 4 and 6, but verify the completeness of the logs: do the logged events capture all relevant information for the logs to be useful for auditing and ultimately accountability purposes?
- 8. For a subset of logs in steps 4 and 6, extract the authenticity data and verify it.

4.2.2 Agent auditing

Our approach to platform auditing is described in Section 5.2.2 of D4.1 [1]. Here, we first provide an overview and then describe how to evaluate the integration of the agent auditing into the PAPAYA platform.

4.2.2.1 Overview

Agents log their processing to standard out, allowing the container environment (such as Docker) to capture them. Quickly viewing agent logs generated by the container running an analytics agent is possible through the Agent Dashboard, see Section 5.2. The use the Agent Dashboard is mainly intended during the development phase of using the PAPAYA platform. During the operational phase, agent logs can use the same Elastic stack as the platform auditing.

4.2.2.2 Integration evaluation

See the Agent Dashboard in Section 5.2 for the development stage. For the re-use of the Elastic components used for platform auditing, we plan to produce a small proof-of-concept where logs generated by the agent is shown in an instance of Kibana running on localhost. Further evaluation is pointless; most PAPAYA users will already be operating some form of logging infrastructure. By showing that the widely used Elastic stack can be used we demonstrate that the PAPAYA platform can be operationalized as part of a larger existing environment.



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4.3 Key Manager

The Key Manager (KM) component is devoted to providing a collateral service of the use of a privacy-preserving platform, carrying out the management of the cryptographic material during the whole lifecycle of the PAPAYA project in the cases where it will be required. In the previous version of this report, it was described the architecture and a detailed design of the Key Manager. Nevertheless, in the following sections of this document are focus on providing the deployment details necessary to use this service and updating that information that have changed since the last report.

4.3.1 Main components and their relationships

The main components and the integration of them within the PAPAYA platform is described in the section 1.1 of the deliverable D4.1 [1].

4.3.2 Deployment and configuration

In order to facilitate the deployment of the Key Manager server, it is available in a docker image. Therefore, to deploy a KM container within the PAPAYA framework it will only be necessary to execute the following command:

\$ docker run -p 9311:9311 --name key-manager -t de.icr.io/papaya-de/key-manager

In addition, the KM client developed to facilitate the integration with other components is available for different programming languages and frameworks, such as: C Sharp, Java, JavaScript, Python, etc.

4.3.3 Implementation constraints

Section 1.1.18 of deliverable D4.1 [1] describes in detail the implementation constraints associated with the implementation of this component.

4.3.4 APIs

Although in the previous version of this report it was detailed the API interfaces used for the Key Manger, some small changes on the interface has been done since them. Hence, the final REST API that will allow to store and to retrieve the cryptographic material is as follows:



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POST/secrets

POST	/secrets Creates a Secret entity
Creates a metadata	a Secret entity. If the payload attribute is not included in the request, then only the for the secret is created, and a subsequent PUT request is required

Parameters		Try it out
Name	Description	
X-Project- ID * required string (header)	X-Project-ID	
<pre>body * required object (body)</pre>	Secret object that needs to be added Example Value Model	
	<pre>{ "name": "string", "expiration": "string", "algorithm": "string", "bit_length": 0, "mode": "string", "payload": "string", "payload_content_type": "application/octet-stream", "payload_content_encoding": "string", "secret_type": "application/octet-stream" }</pre>	
	Parameter content type application/json ∽	



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Response	s Response content type application/json ~	
Code	Description	
201	Successfully created a Secret Example Value Model	
	{ "secret_ref": "string" }	
400	Bad Request	
401	Invalid X-Auth-Token or the token doesn't have permissions to this resource	
403	Forbidden. The user has been authenticated, but is not authorized to create a secret. This can be based on the user's role or the project's quota	
415	Unsupported media-type	

GET/secrets

GET	/secrets Lists a project's secrets
Lists a pro actual sec secret pay	ject's secrets. The list of secrets can be filtered by the parameters passed in via the URL. The ret payload data will not be listed here. Clients must instead make a separate call to retrieve the load data for each individual secret



Parameters	Try it out
Name	Description
X-Project- ID * required	X-Project-ID
string (header)	
OffSet integer (query)	The starting index within the total list of the secrets that you would like to retrieve
	offset - The starting index within the total list
limit integer (query)	The maximum number of records to return (up to 100). The default limit is 10
	limit - The maximum number of records to ret
name string (query)	Selects all secrets with name similar to this value
	name - Selects all secrets with name similar t



alg string (query)	Selects all secrets with algorithm similar to this value
	alg - Selects all secrets with algorithm similar
mode string (query)	Selects all secrets with mode similar to this value
	mode - Selects all secrets with mode similar t
bits integer (query)	Selects all secrets with bit_length equal to this value
	bits - Selects all secrets with bit_length equal
secret_type string (query)	Selects all secrets with secret_type equal to this value
	secret_type - Selects all secrets with secret_t
acl_only boolean (query)	Selects all secrets with an ACL that contains the user. Project scope is ignored



created string (query)	Date filter to select all secrets with created matching the specified criteria. The values for this parameter are comma separated lists of time stamps in ISO 8601 format. The time stamps can be prefixed with any of these comparison operators: gt: (greater-than), gte: (greater-than-or-equal), lt: (less-than), lte: (less-than-or-equal).
	created - Date filter to select all secrets with (
Updated string (query)	Date filter to select all secrets with created matching the specified criteria. The values for this parameter are comma separated lists of time stamps in ISO 8601 format. The time stamps can be prefixed with any of these comparison operators: gt: (greater-than), gte: (greater-than-or-equal), It: (less-than), Ite: (less-than-or-equal).
	updated - Date filter to select all secrets with
expiration string (query)	Date filter to select all secrets with expiration matching the specified criteria. The values for this parameter are comma separated lists of time stamps in ISO 8601 format. The time stamps can be prefixed with any of these comparison operators: gt: (greater-than), gte: (greater-than-or-equal), It: (less-than), Ite: (less-than-or-equal).
	expiration - Date filter to select all secrets wit
SORT string (query)	Determines the sorted order of the returned list. The value of the sort parameter is a comma-separated list of sort keys. Supported sort keys include created, expiration, mode, name, secret_type, status, and updated. Each sort key may also include a direction. Supported directions are :asc for ascending and :desc for descending. The service will use :asc for every key that does not include a direction.
	sort - Determines the sorted order of the retu



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Response	Response content type application/json v
Code	Description
200	Successfull request Example Value Model
	<pre>{ "secrets": [{ "name": "string", "expiration": "string", "algorithm": "string", "bit_length": 0, "mode": "string", "payload_content_type": "application/octet-stream", "payload_content_encoding": "string", "secret_type": "application/octet-stream" }], "total": 0, "next": "string", "previous": "string" } </pre>
401	Invalid X-Auth-Token or the token doesn't have permissions to this resource

GET/secrets/{uuid}





Parameters		Try it out
Name	Description	
X-Project- ID * ^{required}	X-Project-ID	
string (header)		
<pre>uuid * required string (path)</pre>	The uuid that needs to be fetched.	
	uuid - The uuid that needs to be fetched.	



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Response	s Response content type application/json v
Code	Description
200	Successful request
401	Invalid X-Auth-Token or the token doesn't have permissions to this resource Example Value Model
	<pre>{ "status": "string", "created": "string", "updated": "string", "expiration": "string", "algorithm": "string", "bit_length": 0, "mode": "string", "name": "string", "secret_ref": "string", "secret_type": "application/octet-stream", "content_types": "string" } </pre>
404	Not Found
406	Not Acceptable

PUT/secrets/{uuid}





Parameters	Try it out						
Name	Description						
X-Project- ID * required string (header)	X-Project-ID						
<pre>uuid * required string (path)</pre>	secret that need to be added the payload						
	uuid - secret that need to be added the paylo						
<pre>body * required object (body)</pre>	Updated secret object Example Value Model						
	<pre>{ "name": "string", "expiration": "string", "algorithm": "string", "bit_length": 0, "mode": "string", "payload: "string", "payload_content_type": "application/octet-stream", "payload_content_encoding": "string", "secret_type": "application/octet-stream" } Parameter content type text/plain </pre>						
Responses	Response content type text/plain ~						
Code D	escription						
204 S	uccessful request						
401 Ir	Invalid X-Auth-Token or the token doesn't have permissions to this resource						
404 N	Not Found						



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DELETE/secrets/{uuid}

DELETE /secrets/{uuid} Delete a secret by uuid						
Delete a secr	et by uuid.					
Parameters	Try it out					
Name	Description					
X-Project- ID * required string	X-Project-ID					
<pre>(header) uuid * required string (path)</pre>	d The uuid of the secret that needs to be deleted					
	uuid - The uuid of the secret that needs to be					
Responses	Response content type application/json v					
Code	Description					
204	Successful request					
401	Invalid X-Auth-Token or the token doesn't have permissions to this resource					
404	Not Found					



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GET/secrets{uuide}/payload

GET /	secrets/{uuid}/payload Retrieve a secret's payload.						
Retrieve a se	cret's payload.						
Parameters	Try it out						
Name	Description						
X-Project- ID * required	X-Project-ID						
string (header)							
uuid * ^{required} string	The uuid that needs to be fetched.						
(ρατη)	uuid - The uuid that needs to be fetched.						
Responses	Response content type text/plain v						
Code	Description						
200	successful request						
401	Invalid X-Auth-Token or the token doesn't have permissions to this resource						
404	Not Found						
406	Not Acceptable						



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4.3.5 Integration evaluation

This set of components has been designed and implemented to give support to other components and then help them to be integrated within the PAPAYA platform. Therefore, the Key Manager Service will be integrated in some of the use cases whenever necessary. Hence the evaluation of this component will be included with the specific evaluation plans where this set of components will be integrated.



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5 PAPAYA Dashboards

There are two dashboards in PAPAYA, namely the platform dashboard and the agent dashboard. The platform and agent dashboards are, as the name suggests, tied to the respective architectural components and their respective back-ends provide for the dashboards. The dashboards are accessed through web views in a web browser by their respective target users.

5.1 Platform Dashboard

Platform dashboard implemented as a Web application hosted in a container that runs on the PAPAYA's K8s cluster. The major differences, with respect to D4.1, in this section are the implementation constrains (due to internal security constrains, since the platform are deployed on the internal IBM's account). The dashboard will provide the following functionality:

- 1. Present a list of services provided by the platform (Service's Catalog)
- 2. Add/Edit/Delete services.
- 3. Create/Deploy/Delete application (a dedicated instance of each of the provided services).
- 4. Allow application's owners to monitor the flow of the application by presenting operational logs.

5.1.1 Main components and their relationships

As described in the previous deliverable (D4.1 [1]).

5.1.2 Deployment and configuration

As described in the previous deliverable (D4.1 [1]).

5.1.3 Implementation constraints

The dashboard assumes that the user that sends a request is already authenticated (by Identity and Access Management (IAM) mechanism described in Section 4.1).

The platform dashboard supports deployment of three kinds of services:

- Rest server service which communicates via https only
- Rest and TCP server service which communicates via https and socket
- TCP server service which communicates via socket only

Due to internal IBM's security constrains the platform dashboard is limited to 50 TCP channels.

5.1.4 APIs

As described in the previous deliverable D4.1.



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5.1.5 Integration evaluation

We deployed the platform dashboard on the PAPAYA's K8s server. Then we registered a few services and deployed them using the dashboard. In particular, we deployed the service for Privacy-preserving NN classification based on hybrid approach described in Section 3.1.4, and the service for Collaborative Training of Neural Network described in the Section 3.2 (both, the server-side and the client-side), and run the entire flow. In this way we evaluated the functionalities 1-3 listed in the Section 5.1. The evaluation of functionality number 4 (related to presenting operational logs to the application owners) will be presented in D4.3 [6].

5.2 Agent Dashboard

The agent dashboard is briefly described in Section 6.2 of D4.1 [1]. Here, we first provide an overview and then describe how to evaluate the integration of the agent dashboard.

5.2.1 Overview

The agent dashboard is primarily intended to be used during the development phase of a platform client integrating the use of an analytics agent into its systems. There may be some limited use also operationally. The agent dashboard consists of a minimal backend (command line executed) intended to be run on the same system as the agent container and a web-based front-end. Once run, the dashboard opens up a browser interface on localhost that shows the configuration of the run agent and its logs (re-using the visualizations from the Data Subject Toolbox, see Section 6).

5.2.2 Integration evaluation

Do the following steps:

- 1. Download an agent from the PAPAYA platform.
- 2. Start the agent in a Docker container.
- 3. Start the agent dashboard, providing the name of the Docker container from step 2.
- 4. Open the address printed in the terminal in your browser.
- 5. View the configuration of the agent in the browser.
- 6. Perform some analytics with the agent.
- 7. View the logs of the agent in the browser.



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6 Data Subject Toolbox

Data protection by design is a key consideration of any data controller and data processor since the GDPR came into effect. The use of complex cryptographic systems—such as privacy preserving data analytics—pose a challenge for data controllers and processors when dealing with data subjects. In particular, when it comes to the rights to data subjects to be informed about how their personal data is processed and to remain in control of this processing. Towards addressing these challenges, the PAPAYA framework includes a data subject toolbox with a number of small, independent tools that can be used by data controllers and data processors in their interaction with data subjects.

6.1 Explaining Privacy-preserving Analytics

The components for explaining privacy-preserving analytics were described in Section 7.1 of D4.1 [1] as well as in D3.2 [17]. Here, we first provide an overview and then describe how to evaluate the integration of the components.

6.1.1 Overview

The components can be split in two ways. First, there are independent components for explaining risks to data subjects. These components are generic (in the sense that they are applicable for any privacy-preserving analytic) and tied to the output of the CNIL PIA tool¹¹, which we have also improved¹². These components and the modifications are explained in detail in D3.4 [18]. The other type of component aims to explain in easy terms and with usable visualizations how a specific privacy-preserving analytic works. These components are tailored to only one or a particular group of privacy-preserving analytics (e.g., those based on partial or fully homomorphic encryption). These are also explained in detail in D3.4 [18], with a focus on the design of their user interfaces.

Each component mentioned above are independent of each other, enabling a data controller or processor to pick the relevant components for their use of the PAPAYA platform. The components are implemented (some done, others scheduled for M24-M30) as React Native¹³ components. Such components can easily be integrated into existing iOS and Android apps. This should allow the data controller or data processor, as part of a mobile app they provide to data subjects (~users) of their services, to easily integrate the relevant components to explain how the data processing using privacy-preserving analytics works and any associated risks to data subjects.

6.1.2 Integration evaluation

For the components that explain risk:

¹¹ <u>https://www.cnil.fr/en/privacy-impact-assessment-pia</u>

¹² https://github.com/papaya-h2020/pia, modifications released as open source

¹³ <u>https://reactnative.dev/</u>



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- 1. Using the modified CNIL PIA tool, perform a PIA up to and including the risk assessment with respect to the six risk categories in the modified tool.
- 2. Export a JSON-file of the PIA from within the CNIL PIA tool.
- 3. Use the JSON as input to one of the risk components in React Native. Verify that the communicated risk assessments are the same as specified in the PIA from step 1.

The components that explain a privacy-preserving analytic are not integrated with any other component, but rather integrated into an existing mobile app.

6.2 Data Disclosure Visualization Tool

The Data Disclosure Visualization Tool is briefly described in Section 7.2 of D4.1 [1]. Here, we first provide an overview and then describe how to evaluate the integration of the tool in an existing mobile app.

6.2.1 Overview

The goal of the tool is to provide a quick visualization for data subjects of what personal data (by attribute, such as first name, email, and blood type) has been disclosed concerning the data subject to which entities (data processors or data controllers). The design of the visualization is based on the Data Track¹⁴, an open source tool from the EU FP7 project A4Cloud. This visualization should be appropriate for use in a mobile app and will be a single React Native component. The component will be configured by a JSON file that describes the disclosed data attributes and their recipients.

6.2.2 Integration evaluation

There is no integration with other components from PAPAYA. The integration should be evaluated on the ease of integration with an existing mobile app from a developer's perspective.

6.3 Annotated Log View Tool

The Annotated Log View Tool is briefly described in Section 7.3 of D4.1 [1]. Here, we first provide an overview and then describe how to evaluate the integration of the tool.

6.3.1 Overview

The goal of the tool is to make data processing, as it relates to privacy-preserving data analytics, more transparent towards data subjects. The tool consists of two components: one user-interface component and a log annotator component. The log annotator component is used by the data processor or data controller using a platform agent to create annotated logs of its data processing using the agent. The annotated logs are then provided to the user-interface component that visualizes the annotated log. The visualization is based on the "timeline" view of the Data Track,

¹⁴ <u>https://github.com/pylls/datatrack</u>



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but ported to a mobile interface. How the annotated log is stored and provided to the UI component is unspecified, it depends on how the Annotated Log View Tool is used. The implementation of the UI component will be as a single React Native component, and the log annotator component will be provided as a function in several programming languages, including Java and Kotlin for mobile development targeting Android.

6.3.2 Integration evaluation

Verify the integration as follows:

- 1. Setup an analytics agent.
- 2. In the code that instructs the agent to perform some analytics, add a corresponding call to the log annotator component to produce a log entry. Store all log entries in a data structure that maps data subjects to relevant log entries.

In an existing mobile app, integrate the UI component. When the UI component is triggered, supply it with the relevant log entries for the data subject stored in step 2.

6.4 Privacy Engine

6.4.1 Main components and their relationships

The deliverable D3.2 [17] details the design and architecture description of the PE and its subcomponents. No significant changes have occurred on the design or architecture of this components since then. In addition, in the previous version of this document, the relationships between the Privacy Engine components with the rest of the PAPAYA framework components and a full detailed description of the interfaces used by the PE was detailed. Nevertheless, the following sections of this document are focused on the description of the deployment and configuration of the different set of components of the PE.

6.4.2 Deployment and configuration

This section details how to deploy the different sub-components of the Privacy Engine. Following a classical approach these sub-components can be classified into back-end and front-end component classification In addition and just as a reminder, the functionality of the Privacy Engine can be also divided into: Privacy Preferences Manager (PPM) and Data Subject Rights Manager (DSRM). So, the following sections describe how to deploy the Privacy Engine components classify by: Back-End Servers and Front-End Interfaces.

6.4.2.1 Back-End Servers

The back-end components have been implemented to be containered and therefore, easily deployed using dockers. As mentioned before, two different images have been developed, one for the PPM functionality and other for the DSRM. For the PPM deployment it will be necessary to execute the following command:

\$ docker run -p 8080:8080 de.icr.io/papaya-de/privacy_engine-ppm-server:latest



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While for deploying a DSRM container, it will be as follows: \$ docker run -p 8080:8080 de.icr.io/papaya-de/privacy_engine-dsrm-server:latest

6.4.2.2 Front-End Interfaces

The front-end interfaces have been developed to be included in mobile applications. In order to maximize the possible platforms where these interfaces can be integrated, they have been developed using the NativeScript framework¹⁵. The use of this popular and broadly used framework facilitates not only the development but also the integration within the final applications used for the PAPAYA pilots. How to integrate these interfaces with the final pilots applications can be consulted in the following link:

https://www.nativescript.org/faq/how-do-i-add-nativescript-to-an-existing-ios-or-android-app

6.4.3 Implementation constraints

There are no further implementation constraints since this was reported in the previous version of this document.

6.4.4 APIs

The whole API of the different components of the Privacy Engine can be consulted in the deliverable D3.2 [17].

6.4.5 Integration evaluation

The Privacy Engine will be included in two different Use Cases: US2 - Privacy-preserving stress management and US3 - Privacy-preserving mobility analytics. Thereby PE evaluation will be included in the plan developed for the evaluation of both Use Cases.

¹⁵ <u>https://www.nativescript.org/</u>



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7 Platform Deployment

There are minor changes to the deployment process (in comparison to D4.1 [1]). In order to make the description self-contained we describe the whole deployment process in this section.

7.1 Platform initialization and platform dashboard deployment

For the proof of concept (PoC) usage, the platform is deployed on IBM's cloud service. The platform administrator (IBM) created an account of K8s service and allocated the required resources for the Kubernetes cluster. In addition, the platform administrator allocated the following required external services:

- 1. Container Registry (CR). The platform administrator created single namespaces, papaya-de.
- 2. Permissions Provided to the platform users
- 3. Elasticsearch and Kibana internal instance.

The platform administrator created the following permission groups:

- 1. Platform administrator
 - a. Allocate resources on the Kubernetes cluster
 - b. Allocate/Create external services
 - c. Admin access to CR
- 2. Service providers
 - a. Active access to the CR
 - b. If the service requires integration/communication with an external service, appropriate accounts should be created or allocated.
- 3. Platform Clients
 - a. Should be able to download only the agent side container from the CR

The platform administrator deployed the Platform Dashboard using yaml deployment file. Service providers and platform clients can use it to deploy/use the services.

To support auditing, the platform administrator will execute Loggingbeat as Deamonset on the Kubernetes cluster and connect it to the Elasticsearch instance.

7.2 Service upload and deployment

This section describes the steps needed in order to upload or use a service. For the PoC usage we use a Container Registry Provided by IBM Cloud. The steps are divided into two phases: (1) uploading phase; and (2) usage-execution phase.



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The following steps are common for both phases:

- 1. Install the IBM Cloud CLI¹⁶
- 2. Install the Docker CLI ¹⁷
- 3. Install the Container Registry plug-in. ibmcloud plugin install container-registry -r Bluemix
- Log in to your IBM Cloud account. ibmcloud login -a https://cloud.ibm.com

The uploading phase will be done by the service provider user and it consists of the following steps:

- 1. Log your local Docker deamon into the IBM Cloud CR ibmcloud cr login
- 2. Create two images: server side and client-side agent. The server-side container will run on papaya's K8s cluster. The client-side agent will run on the client's side. Instructions on how to create the Docker image can be found in the following link:

https://docs.docker.com/develop/develop-images/baseimages/

3. Upload each image to the appropriate namespace on the CR for the service images use:

docker	tag	<local< th=""><th>container</th><th>name></th><th>de.icr.io/papaya-</th><th>de/<my_< th=""><th>_repository>:<my< th=""><th>_tag></th></my<></th></my_<></th></local<>	container	name>	de.icr.io/papaya-	de/ <my_< th=""><th>_repository>:<my< th=""><th>_tag></th></my<></th></my_<>	_repository>: <my< th=""><th>_tag></th></my<>	_tag>
docker		push	de.icr.	io/	papaya-de	/ <my_< td=""><td>_repository>:<my< td=""><td>_tag></td></my<></td></my_<>	_repository>: <my< td=""><td>_tag></td></my<>	_tag>

More details and CR commands can be found in Kubernetes documentation¹⁸.

4. The service provider user (i.e. IBM, ORANGE, EUROCOM) will add a new service to the services catalog on the Platform dashboard (https://platform.papaya.eu-de.containers.appdomain.cloud), while specifying all required fields and describing in details the service's flow and requirements.

The usage-execution flow will consist of the following steps:

- 1. The platform client will login to the Platform dashboard (https://platform.papaya.eu-de.containers.appdomain.cloud/auth/).
- 2. The client will select the tab of services catalog and choose the service of interest.
- 3. The client will provide a configuration file if the service requires so.
- 4. The client will download the client-side agent image by using the following commands:

¹⁶ https://cloud.ibm.com/docs/cli?topic=cloud-cli-getting-started#idt-prereq

¹⁷ https://docs.docker.com/install/

¹⁸ https://cloud.ibm.com/kubernetes/registry/main/start



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ibmcloud cr login docker pull <image name>

Where the <image name> is presented in the "services catalog" view on a platform dashboard under the "Agent Side Container" name and will be structured of the following:

de.icr.io/papaya-de/<image>:<tag>

5. After the services are deployed on the k8s cluster (by the platform dashboard application), the platform will present the url of the deployed service for the https communication or provide a cluster IP and port for TCP (socket) communication, For example:

SERVER_URL=https://29dr4d.papaya.eu-de.containers.appdomain.cloud SERVER_TCP_PORT=32067 SERVER_IP=158.177.182.219

In addition, the platform dashboard provides the option to download the file (env file) that will contain these parameters (accordingly to the communication protocol with the service). The client administrator should be able to provide this file to the client agent, based on how the agent expects to receive these parameters. Possible ways of doing that are as follows:

a. Provide the env file as env. variables for the client agent execution.

docker run with an --env-file <file name>

b. Read the env file by the client's application and provide it to the agent side within the initiation step. Meaning that the agent side will provide INIT API and will expect to receive all essential parameters in order to communicate with appropriate server side.



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8 Conclusions

In this deliverable, we presented modifications and extensions to the platform functional design, architecture and deployment presented in D4.1 [1]. Thus, this deliverable completes the specification presented in D4.1 [1]. We described in detail the core platform services dealing with privacy-preserving computations as well as the services responsible to ensure data privacy, security, and transparency of all the processes while operating the platform. We explained how different services can be integrated into the platform in a way that they will be interoperable/compatible with each other and could work together in the integrated platform. In particular, by using k8s, we show that the analytics are possible to run using modern cloud environments; by adding Identity and Access management (IAM), we ensure that access to the analytics can be properly authenticated and authorized; by adding auditing mechanisms, we ensure that the analytics generate appropriate logging information.

We also presented the design of platform dashboards that will provide UI, configuration, and visualization functionality. Finally, we described how we deployed some of the services and how we evaluated their integration.

In next version of this document (D4.3 [6]) we plan to provide the full description of incomplete services, missing evaluation results, and changes to the current design that maybe be required after the platform evaluation in WP5.


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