# Modulus: An Open Modular Design for Interoperable and Reusable Machine Learning

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# Abstract

Modulus provides an open framework for developers to create modular, interoperable modules. These modules are designed to be modular (surprise?), reusable and inter-operable locally and remotely via peer to peer communication protocols. Modules are lightweight and general enough to wrap over any machine learning tool. Developers can also organize modules into a module file system, representing their own module hub. Developers can also expose their modules as public endpoints through their local peer, and can restrict access based on their accounts signature. Modulus is by design open source and does not rely on any tokenomics, allowing developers to monetize their public endpoints through any tokenized asset including their own.

# 1 Introduction

Sharing is caring, especially in machine learning. However, many of the models and pipelines are not very reusable, requiring additional tedious effort towards interoperability. This lack of interoperability allowed for platforms fill this void (ZenML, Lightning, Fast.ai, Huggingface). These companies attempt to connect different tools into one cohesive tool-set, striving to be the one platform to rule them all. Despite many platforms proclaiming their support for open ML, many do not fully open source their code-base, and some monetize their community's intellectual property to cooperate clients . Platforms can also lock-in developers, as they build tools that are fully compatible with their ecosystem and not competitors. Paradoxically, these platforms create walled gardens that section developers from effectively collaborating. Modulus attempts replace walled gardens with a wall-less garden; a fully open source protocol for developers to create, connect and share machine learning modules . Modulus provides modules templates, allowing tools to be reusable, inter-operable, and connectable via local or remote messaging protocols (TCP, UDP, gRPC). Developers are provided with their own module hub, where they can organize share and monetize modules.

# 2 Methods

Modulus is strongly opinionated in being un-opinionated. The design attempts to serve as an inter operable layer that allows users to easily wrap over any machine learning or processing with its module block.

## 2.1 Module Block

A module block is a process that allows for multiple inputs and outputs. It can receive inputs and write outputs via its set of clients which connect to various servers/processes outside of the module. These servers involve local/remote databases (Redis, MongoDB, Postgres), API endpoints

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(GraphQL/REST), peer to peer connections (gRPC[6], IPFS[1], Arweave[14]) and customised connections. A typical module run-time involves three steps: 1. **read** preliminary data from a clients (loading a dataset, model) 2. **run** some custom logic over the read data into output data 3. **write** output data to clients. If the read, run, write flow is too limiting, users can build their own custom methods as they would with a normal class.



Figure 1: Illustration of a module's forward pass

# 3 Module File-system

Each module is represented as a folder. The folder contains the main python script (module.py) and its configuration file (module.yaml). This structure allows modules to be inter-operable with practically any asset that can be represented as these two files. For instance, you can wrap a Huggingface model/Dataset with a module, and feed it with tweets from a model wrapped over a Twitter API. The module folder can also have additional files and dependencies. This can include other modules (folders), Python files, Dockerfiles, pip/conda environment files, Helm charts, . These files help ensure that the module is portable wherein its run-time can be compatible with multiple machines.



Figure 2: module tree containing a file-system of modules.

Module folders can be constructed into a single file system representing an ontology/tree of modules. The tree helps developers organize modules based on their purpose and classification (model, dataset, process). The module tree can also be synonymous to a mono-repo, and is intended to contain modules

serving across multiple projects. This encourages developers to have their own personal module tree mono-repo, where they can easily reuse and share modules for new projects.

#### 3.1 Module Manager API

The module manager allows developers to manage running modules and submit tasks to these modules. The module manager is also responsible for monitoring and managing the modules' resource consumption. At the moment, the total module consumption is tracked in terms of disk used, number of CPUs, main memory and GPU memory. When a new module needs to replace an existing module, the outgoing module marked with the lowest priority score. This priority score is defined by the user, and defaults to first in first out (FIFO). Launched modules are also provided with public/private key pairs. These module keys are used for assigning each module an identity that is verified with their private key signature. Modulus also uses multi-format schemes to accommodate the verification and identity that is agnostic to the key type (ie, ECDSA [11], AES[2]). This allows for modules to have identities as account on multiple block-chains, having a key pair for each chain. This identity layer can also limit external peers from accessing publicly exposed modules.

#### 3.2 Launching a Module

A module can be initialized by knowing its module path. If no configuration is specified, the module's default YAML file is used (module.yaml). Users can also override the default configuration with an override dictionary mapping the configuration key path ('dataset.batchsize'=12) with the replacing value. Modules can be initialized as ray actors [9]. Ray actors are class objects that can be deployed as daemon processes over a a local or Kubernetes cluster. When deployed, each module actor is given a name based on its module path. If more than one of the same module is deployed (ie. a pool of actors), a tag is used to avoid name conflicts (model.transformer-1 and model.transformer-2). These processes live until they are killed (ray.kill) allowing for developers to reuse already deployed modules as remote processes.

#### 3.3 Calling a Module

Jobs are submitted through a REST API giving access to the module manager. The user can input a JSON serialized python dictionary involving 3 key components: **1. The module's path (string), 2. the function attribute (string) , and 3. The functions inputs (list/args or dictionary/kwargs)**. If the module is not running in a ray cluster, a new module will be spawned.



Figure 3: Illustration of API call.

#### 3.4 Connecting Modules

Connecting modules occurs locally or over the wire with modules located on an external peer. Messaging between peers occur through the queue server.

#### 3.5 Queue Server

The queue server is a special module that is responsible for managing a dictionary of Asyncio queues [7]. This allows for modules to access each other through putting and getting objects from a global

queue. By default, each module has an input and output queue, where the queue's key is defined by its module path, suffixed by the queue direction. For instance, module "**model.transformer**" would have an input and output queue under "**model.transformer.in**" and "**model.transformer.out**". This is not strict and modules can create their own custom names for queues for specific use cases. However, allowing for a default naming of queues makes it simple for developers to chain modules knowing their names.



Figure 4: Illustration of a Queue Server

# 4 Module Pipeline

Queue servers connect modules through sending and receiving objects in their key registered queue. An example of this can be one or many dataset modules sending training batches to a queue for a model module to asynchronously receive and train on. Another example could be having an aggregator model receive the predictions of many models for assembling. Allowing for modules to synchronize on queue's makes it simple for modules to inter operate in many complex forms.



Figure 5: Illustration of a general pipeline of interconnecting modules.

### 4.1 Connecting Modules over the Wire

The modulus API allows for external users to run module actors locally over JSON gRPC [6]. This involves **user A** sending a request to **user B's** machine, which is placed in a request queue before it is processed by the module. Some connections may require authentication signatures from requesting peers. This allows for the server to limit access to accounts they trust. Upon verification, the module returns a response containing a JSON serialize response.

## 4.2 Module Access Control

Module access control allows developers to expose their module API to trusted accounts. These accounts can be collaborators, friends, family and patrons who pay for access. Verification is done through having the client peer sign a string containing a UTC timestamp and its public key address, with the key type. This multi format is used to accommodate for multiple key types for the purpose of agnosticism [5]. if the message was signed by a trusted peer, the server should be able to proceed the API call knowing its a trustful client. This form of control allows developers to monetize their endpoints to patrons. Developers can create custom payment schema that fit their use-case, and can charge in whatever token they desire. This avoids developers from being locked in to any

specific tokenomic structure, giving full financial autonomy. One consideration is that peers may not authenticate frequently to avoid network costs. Each pair of peers can decide their own verification depending on their needs. This can be involve some situations that require more security to have more frequent verification, with the frequency of verification's drop proportionally to the level of trust between two peers.



Figure 6: Authentication process in a server module verifying a client module.

## 4.3 Smart Contracts

Modulus uses python libraries that interface with different block-chains. These include web3(EVM) [4], substrate (Polkadot) [10], CosmosSDK [3], and solona-py (Solona) [12]. These smart contract interfaces can be wrapped by a module, allowing for smart contracts to interact with other non-smart contract modules . For example, one module can predict the price of Ethereum, send its prediction to a module infer-facing with a smart contract to trigger to swap Ethereum for USDC on Sushiswap's decentralized exchange. Modules can also connect different smart contracts across different chains, allowing for customize cross chain synchronization for python developers. We refer to such a process as smart contract portals, as it involves building a portal synchronizing two smart contracts. This can be synchronized using one module, or many modules across different peers. Each synchronization may require multi-signature aggregation before triggering a synchronization state change [8]. In the limit, the entire state of the smart contracts can be mirrored across multiple chains.

#### Single-Chain



Figure 7: Smart contract interfacing modules

#### 4.4 Storing Modules

Recall each module is self-contained inside a folder, containing all of the necessary information for portability across peers. The folder can be made publicly accessible through IPFS, Arweave, Ceramic, and Filecoin. Private assets can be encrypted by a private key that is ideally only distributed to provisioned users who can decrypt the files. This ability to limit accessibility allows users to monetize their modules as intellectual property. This can include selling assets, period subscription access , token per call, or profit sharing [13]. In the subscription access example, this involves the provider revealing a private key for registered users, but expires under specific conditions (ie after 30 days). Because the modules are token agnostic, developers can choose their preferred tokens of payment, whether its their a stable coin, Ethereum, or their own token.

# 5 Conclusion

Modulus is a framework that strives towards ensuring users can easily inter-operate and monetize their assets. The framework is intended to provide full autonomy and intellectual property over developer assets. This design avoids developers from succumbing to platform lock-ins, breaking through the walled gardens.

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