

# Containerisation

Sprint 2 - Week 4

*INFO 9023 - Machine Learning Systems Design*

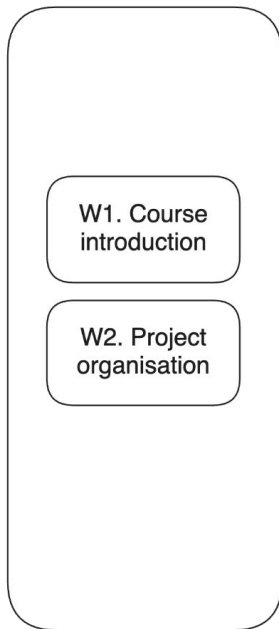
Thomas Vrancken ([t.vrancken@uliege.be](mailto:t.vrancken@uliege.be))

Matthias Pirlet ([matthias.pirlet@uliege.be](mailto:matthias.pirlet@uliege.be))

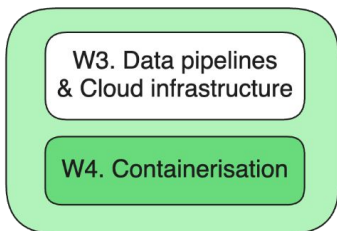
2025 Spring

# Status on our overall course roadmap

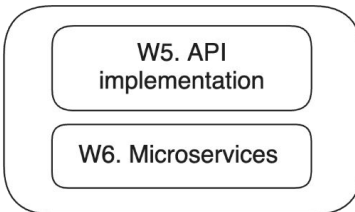
Sprint 1:  
Project organisation



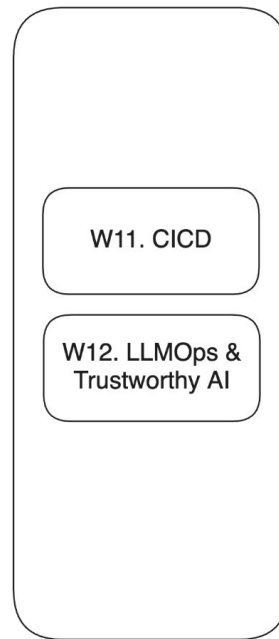
Sprint 2:  
Cloud & containerisation



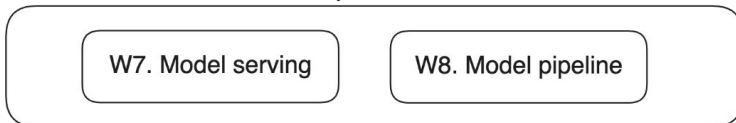
Sprint 3:  
API implementation



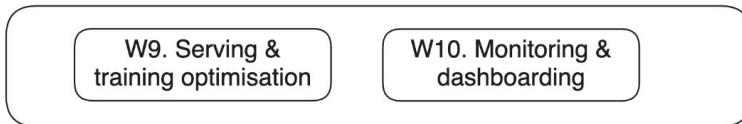
Sprint 6:  
CICD



Sprint 4: Model deployment



Sprint 5: Optimisation & monitoring





# Agenda

What will we talk about today

## Demo

1. Google Cloud Storage & Big Query

## Lecture

2. Cloud infrastructure (deeper dive)
3. Virtual environments
4. Virtual machines
5. Containers

## Directed Work

6. Docker

# Project objective for sprint 2

This course focuses on what comes **around** your ML model.

Do **not** spend significant time optimising your data or model.

**No grading** on the **accuracy** of the model itself.

| #   | Week      | Work package   | Requirement |
|-----|-----------|--|-------------|
| 2.1 | W03       | Prepare your data and run an Exploratory Data Analysis.  | Required    |
| 2.2 | W03       | Prepare your Cloud environment. That means creating a Cloud project, granting correct access rights to all members of your group and setting up a billing account.<br><b>Attention:</b> You can have free credits for the Cloud, as explained during the course. | Required    |
| 2.3 | W04       | Train your ML model  | Required    |
| 2.4 | W04       | Evaluate your ML model   | Required    |
| 2.5 | W03 & W04 | Document your data analysis and model performance  | Required    |

# Tour of Google Cloud

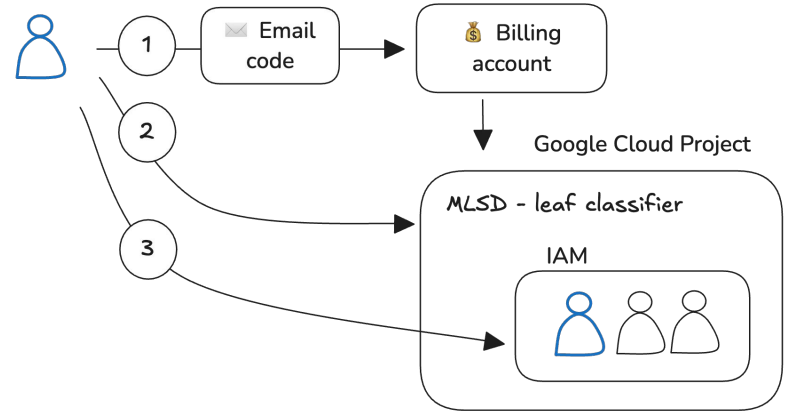
# Cloud & credits

## How to activate credits and manage projects

Only **one student** per group needs to create a project and activate the credits. If you run out of credits you can use the ones of someone else on your team.

Steps to setting up a project

1. Click on the link received by email to redeem credits and attach them to a [billing account](#)
2. Create a Google Cloud [Project](#) and link it to the billing account you created
3. Grant access to the other members of your team to this project through the [IAM portal](#)



⚠ We will do a tour of google cloud and we will demo how to setup a google cloud project!  
No action needed from your side till then.

# Demo: Google Cloud

Heading to Github 👉 [https://github.com/ThomasVrancken/info9023-mlops/blob/main/demos/04\\_gcs\\_bq/README.md](https://github.com/ThomasVrancken/info9023-mlops/blob/main/demos/04_gcs_bq/README.md)



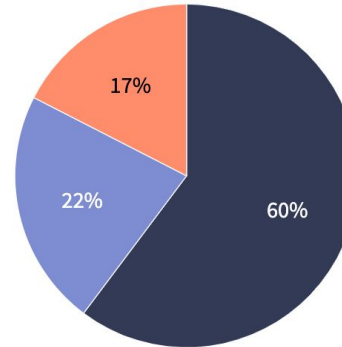
# Cloud infrastructure (deeper dive)

# Cloud is a profitable business.

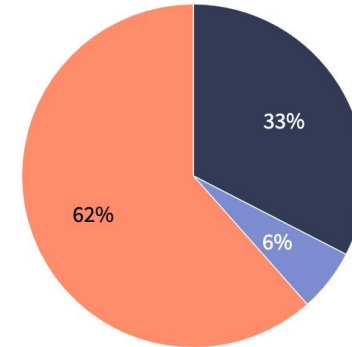
## Amazon Segment Breakdown

Data as of Q2 FY 2024, ended March 31, 2024.

■ North America  
■ International  
■ AWS



Revenue



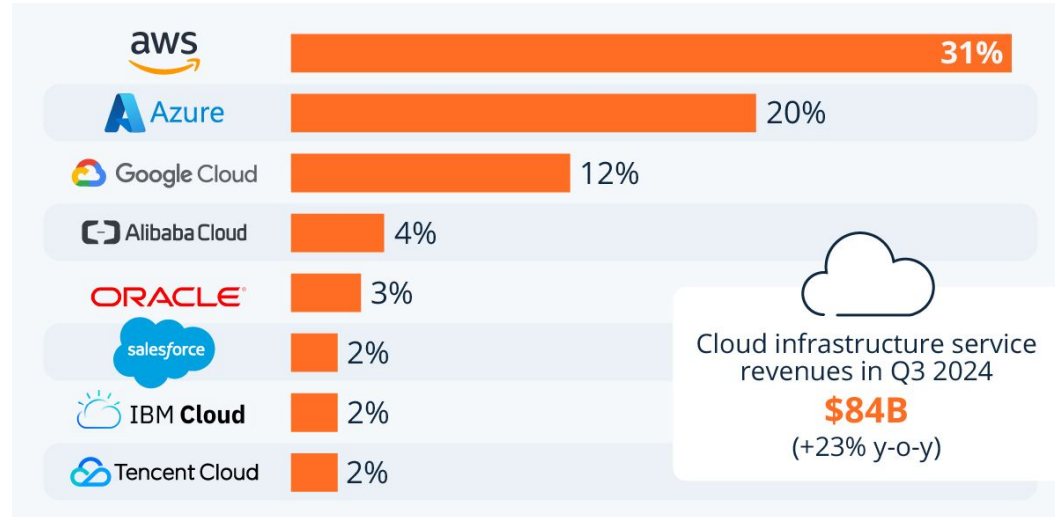
Operating Income

*Percentages may not add to 100% due to rounding.*

Chart: Matthew Johnston • Source: [Amazon 10-Q](#)

 Investopedia

# Usual suspects of the “Cloud”



# Usual suspects of the “Cloud”

Cloud provider offer really similar services.



# Overview of services types on the Cloud

1. **Infrastructure** as a Service (IaaS)
2. **Platform** as a Service (PaaS)
3. **Software** as a Service (SaaS)
4. **Function** as a Service (FaaS) (or Serverless)

# Infrastructure as a Service (IaaS)

On-demand “Pay-as-you-go” **data**, **compute** and **networking** infrastructure.

| Service             | Provider | Type         | Description  |
|---------------------|----------|--------------|--|
| AWS S3              | AWS      | Data storage | Raw (blob) <b>storage</b> infrastructure to store flexible files (documents, images, videos, etc.). Accessible by applications (python sdk). Used to store large amounts of unstructured data (e.g. logs) or pass it between components of your application.                 |
| Cloud Storage (GCS) | Google   |              |  |
| Azure Blob Storage  | Azure    |              |  |
| AWS EC2             | AWS      | Compute      | Provide <b>virtual machines</b> that let you run applications and workloads in the cloud. Give you control over the operating system, software, and configuration while handling the physical hardware and infrastructure.<br>You can SSH into the VM and run what you want. |
| Compute Engine      | Google   |              |  |
| Azure VMs           | Azure    |              |  |

# Platform as a Service (PaaS)

Delivers and *manages* hardware and software resources for developing, testing, delivering and managing cloud applications.

## Compute

| Service                           | Provider | Type                   | Description   |
|-----------------------------------|----------|------------------------|---|
| Elastic Beanstalk                 | AWS      | Application deployment | Build and host web apps, APIs, and mobile backends. Often used for front-end applications. Infrastructure is managed by Cloud provider.   |
| App Engine                        | Google   |                        |   |
| App Service                       | Azure    |                        |   |
| Elastic Kubernetes Services (EKS) | AWS      | Kubernetes hosting     | Cloud-managed kubernetes (k8s) platform. Cloud provider hosts a kubernetes server, by opposition to hosting it on a private server or on-premise compute. Microservices can then be implemented and deployed on the kubernetes servers. The kubernetes server will then manage the scaling of the services. |
| Google Kubernetes Engine (GKE)    | Google   |                        |   |
| Azure Kubernetes Service (AKS)    | Azure    |                        |   |

# Platform as a Service (PaaS)

## Data

| Service           | Provider | Type                              | Description  |
|-------------------|----------|-----------------------------------|--|
| Redshift          | AWS      | Analytical relational database    | Analytical database used to store and run aggregation operations on those data. Often used in ML applications due to the need to perform analytics.  |
| BigQuery          | Google   |                                   |  |
| Synapse Analytics | Azure    |                                   |  |
| RDS               | AWS      | Transactional relational database | Fixed-schema structured database. Used for transactions rather than analytics. Working with relationship tables (fact and dimension) and SQL.  |
| Cloud SQL         | Google   |                                   |  |
| SQL Database      | Azure    |                                   |  |
| DynamoDB          | AWS      | NoSQL databases                   | Document and key-value data model services. Designed for high availability, scalability, and flexible data structures. Support semi-structured (JSON) data and offer low-latency, globally distributed access. |
| Firestore         | Google   |                                   |  |
| Cosmos DB         | Azure    |                                   |  |



# Platform as a Service (PaaS)

## Machine Learning

| Service   | Provider | Type         | Description  |
|-----------|----------|--------------|--|
| Bedrock   | AWS      | ML Platforms | <p>Centralised platform for the implementation of ML models.<br/>Centered around the following offerings</p> <ul style="list-style-type: none"><li>• Feature engineering</li><li>• Model development<ul style="list-style-type: none"><li>◦ E.g. “workbenches” to access notebooks on GPU enabled notebooks</li></ul></li><li>• Model experimentation</li><li>• Model serving &amp; deployment</li><li>• ML pipelines</li><li>• Model monitoring</li></ul> |
| Vertex    | Google   |              |  |
| ML Studio | Azure    |              |  |

*Buzzword list of this course...*

# Software as a Service (SaaS)

Mostly refers to software hosted on the Cloud.

Arguably, some data visualisation tools are SaaS.

| Service  | Provider | Type         | Description  |
|----------|----------|--------------|--|
| Tableau  | AWS      | Dashboarding | Build dashboards and data visuals directly on databases hosted in the Cloud (or imports such as xlsx).<br>Low-code.<br>Facilitate business intelligence.<br><i>Realm of data analysts.</i> |
| Looker   | Google   |              |  |
| Power BI | Azure    |              |  |

# Function as a Service (SaaS) or serverless

*Covered in week 06 of this course*

| Service             | Provider | Type               | Description   |
|---------------------|----------|--------------------|---|
| Fargate (or Lambda) | AWS      | Serverless compute | Fully managed, serverless hosting of microservices. Host <u>containerised</u> applications (mostly APIs). Automatically scales the application based on demand and abstracts all infrastructure management. |
| Cloud Run           | Google   |                    |   |
| Container Apps      | Azure    |                    |   |
| Lambda              | AWS      | Function compute   | Run small isolated pieces of codes (functions). Typically event based. So triggered or scheduled.   |
| Functions           | Google   |                    |   |
| Functions           | Azure    |                    |   |

*Not a typo, just unoriginal naming...*

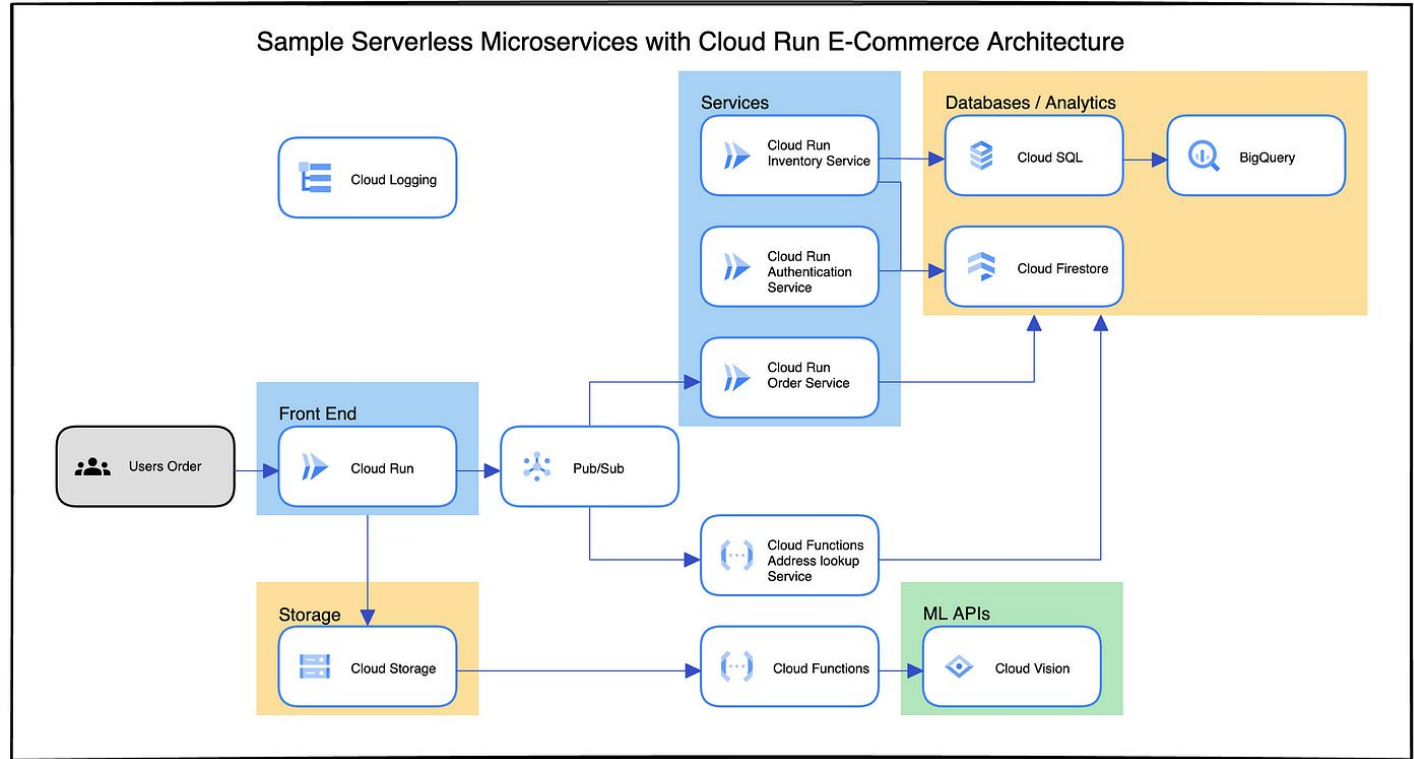
# Out-of-the-box AI models.

Cloud providers also offer pre-trained AI models, such as:

- LLMs (Azure Open AI, Gemini...)
- Document AI
- Optical Character Recognition (OCR)
- Translation
- Speech-to-Text
- ...

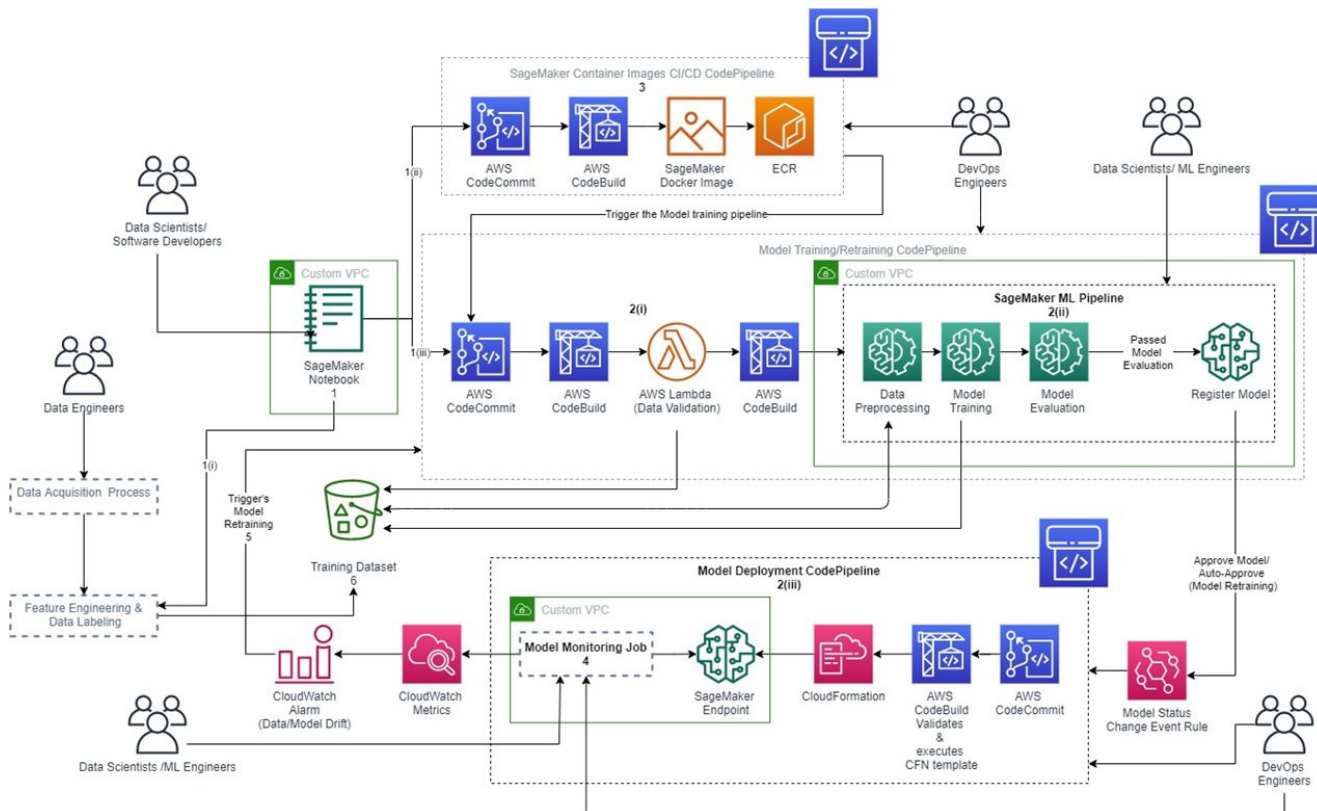
# Example application architecture

Google Cloud



# Example application architecture

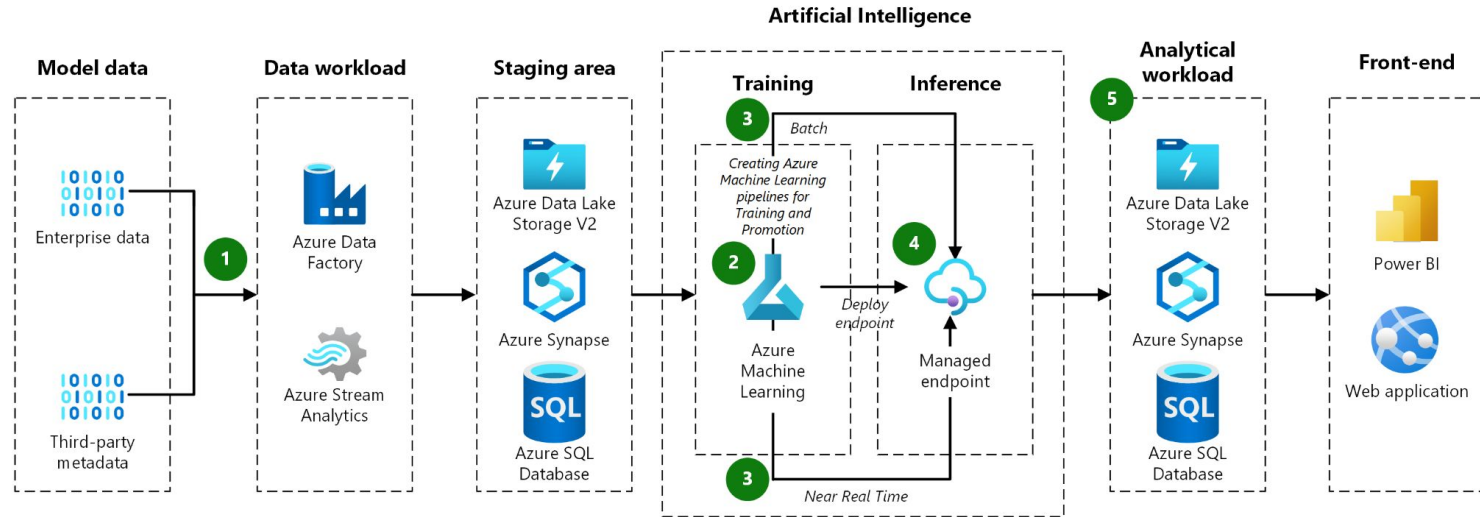
AWS



<https://aws.amazon.com/blogs/apn/taming-machine-learning-on-aws-with-mlops-a-reference-architecture/>

# Example application architecture

AWS



# Virtual environments



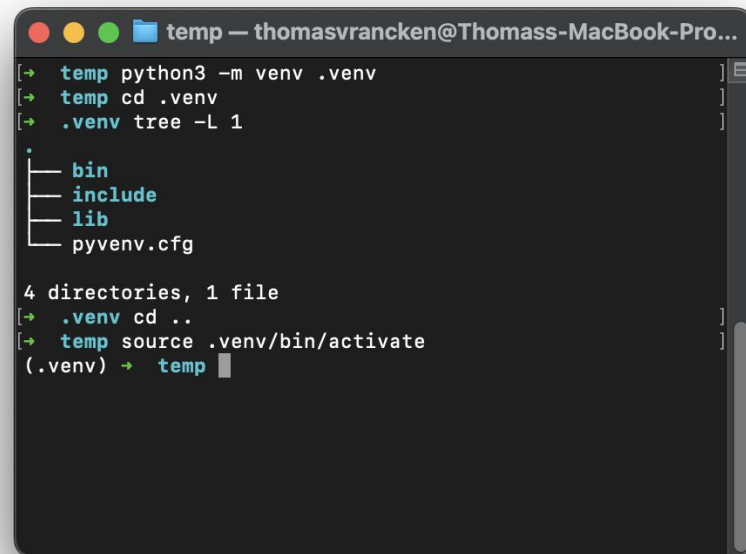
# What is a virtual environment?

- Virtual environments keep **dependencies** (e.g. libraries...) in separate “**environments**” so you can switch between both applications easily with totally different packages/versions.
- Given an operating system and hardware, you can set **different environments** using **different technologies**.
- Virtual environments help to make **development** and use of code more **streamlined on local machines**.
- You might use **different dependencies** (e.g. library/package versions) for **different projects**. Abstract away the dependencies by using a virtual environment.
- It will help maintaining dependencies (e.g. `requirements.txt`)

# What is a virtual environment?

Concretely, a virtual environment is a directory with the following components:

- **Directory** where third-party libraries are installed
- **Links** to the executables on your system (python itself or pip)
- **Scripts** that ensure that the code uses the interpreter and site packages in the virtual environment



```
temp — thomasvrancken@Thomass-MacBook-Pro...  
[→ temp python3 -m venv .venv  
[→ temp cd .venv  
[→ .venv tree -L 1  
.  
├── bin  
├── include  
├── lib  
└── pyvenv.cfg  
  
4 directories, 1 file  
[→ .venv cd ..  
[→ temp source .venv/bin/activate  
(.venv) → temp
```

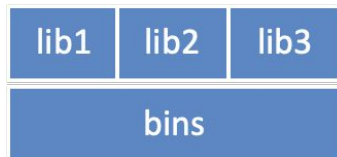
# What is a virtual environment?

Here's what happens when using the virtual environment:

1. **Activation:** When you activate the virtual environment, your shell's PATH is updated to prioritize this bin (or Scripts) directory. This means that when you type python or pip, your shell will use the versions in the virtual environment instead of the system-wide versions.
2. **Execution:** Because of the updated PATH, when you execute Python or pip, your system uses the linked executables in the virtual environment directory. This ensures all Python operations are limited to the virtual environment.
3. **Isolation:** Since these executables are specific to the virtual environment, any Python packages you install or remove affect only this isolated environment, leaving other environments and the system-wide settings untouched.

# Why should you use virtual environments?

Maggie took “Intro to Machine Learning”. She used to run her Jupyter notebooks from anaconda prompt. Every time she installed a module it was placed in the either of `bin`, `lib`, `share`, `include` folders and she could import it in and used it without any issue.



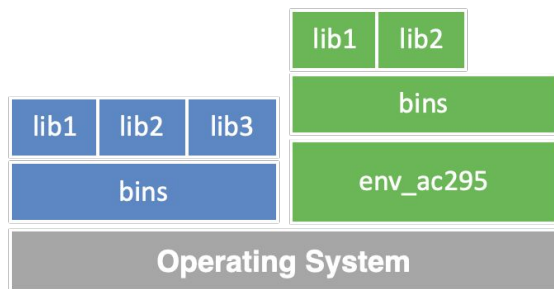
Operating System

**Maggie**

```
$ which python
/c/Users/maggie/Anaconda3/python
```

# Why should you use virtual environments?

Maggie starts taking “Machine Learning Systems Design”, and she thinks that it would be good to isolate the new environment from the previous environments avoiding any conflict with the installed packages. She adds a virtual environment that helps her keep the modules organized and avoid misbehaviors while developing a new project.

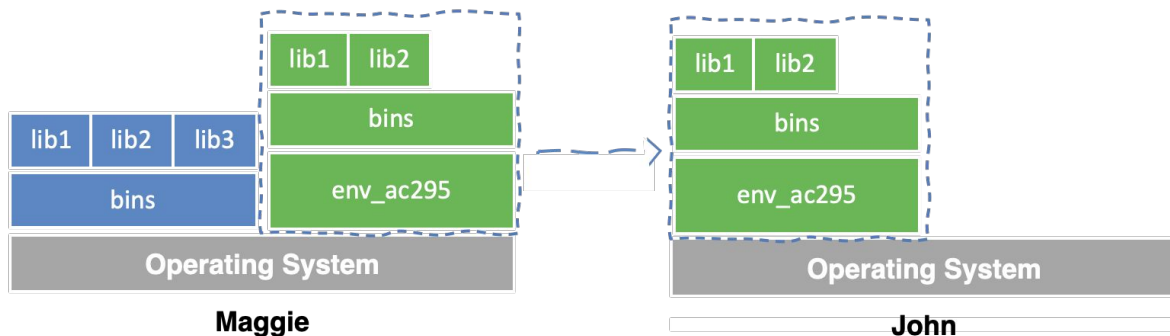


**Maggie**

```
$ which python  
/c/Users/maggie/Anaconda3/envs/env_ac295/python
```

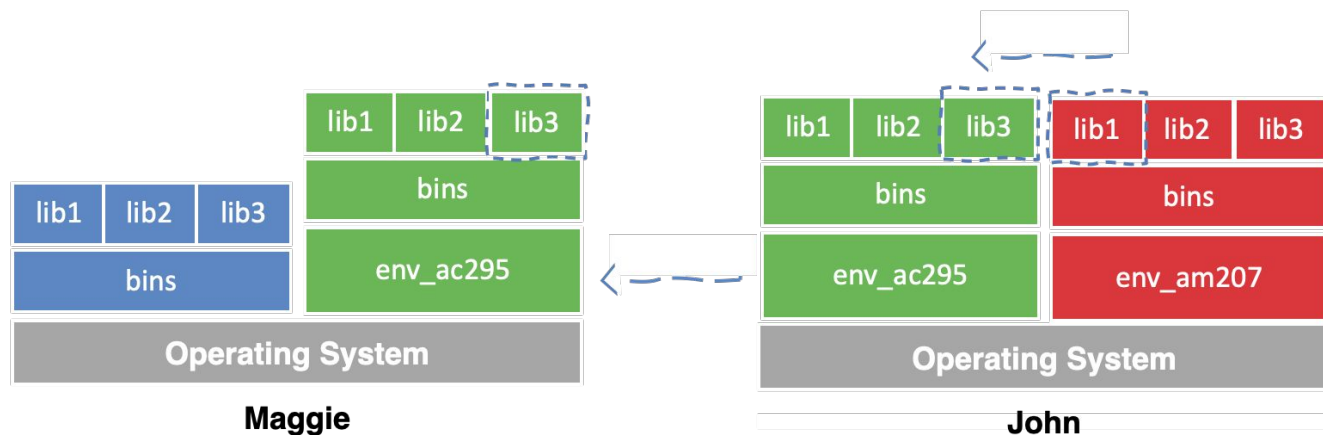
# Why should you use virtual environments?

Maggie collaborates with John for the final project and shares the environment she is working on through `requirements.txt` file.



# Why should you use virtual environments?

John experiments a new method he learned in another class and adds a new library to the working environment. After seeing tremendous improvements, he sends Maggie back his code and a new `requirements.txt` file. She can now update her environment and replicate the experiment.



# Virtual environments: virtualenv vs conda

## virtualenv

- virtual environments manager embedded in Python
- incorporated into broader tools such as pipenv
- allow to install modules using pip package manager

## How to use virtualenv?

- create an environment within your project folder: `python -m venv your_env_name` (often `.venv`)
- it will add a folder called `your_env_name` in your project directory
- activate environment: `source env/bin/activate`
- install requirements using: `pip install package_name=version`
- deactivate environment once done: `deactivate`



# Virtual environments: virtualenv vs conda

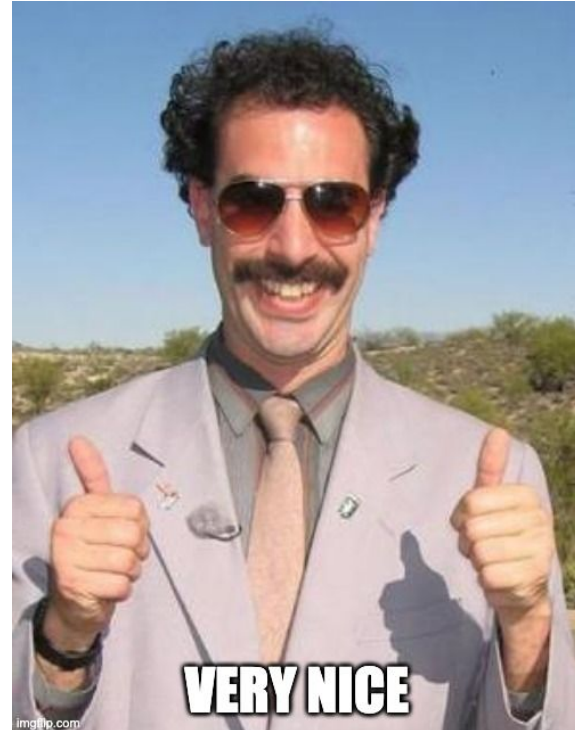
## conda **environment**

- virtual environments manager embedded in Anaconda
- allow to use both conda and pip to manage and install packages

## How to use `conda` ?

- create an environment: `conda create --name your_env_name python=3.12`
- it will add a folder located within your anaconda installation: `/Users/your_username/anaconda3/envs/your_env_name`
- activate environment: `conda activate your_env_name` (should appear in your shell)
- install requirements using: `conda install package_name=version`
- deactivate environment once done: `conda deactivate`
- duplicate your environment using YAML file: `conda env export > my_environment.yml`
- to recreate the environment now use: `conda env create -f environment.yml`
- find which environment you are using : `conda env list`

When you installed all your dependencies only using  
your requirements.txt file



# Why should you use virtual environments?

## Pros

- Reproducible research
- Dependency management (forces you to maintain a requirements.txt with all dependencies)
  - Natural first step before containerisation
- Improved engineering collaboration
- Broader skill set

## Cons

- Effort setting up your environment
- Storage space (duplicated binaries and libraries)
- Tool / OS compatibility (some IDEs and OS won't be able to share the same .venv directory)

# The .gitignore file



```
# These are some examples of commonly
# ignored file patterns.

# Compiled Python bytecode
*.py[co]

# Log files
*.log

# Environmen
*.env
venv/

# Data
data/

# Secrets
secrets/

# Jupyter Notebook
.ipynb_checkpoints
```

# Virtual Machines (VM)

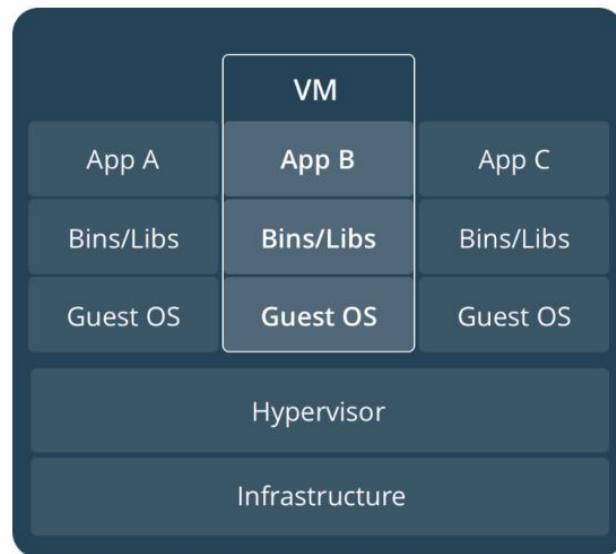
# Virtual Machine (VM)

Provides a fully functional snapshot of an operating system (OS).

Connect to the VM in a similar way as you would connect to a specific computer.

## Motivation

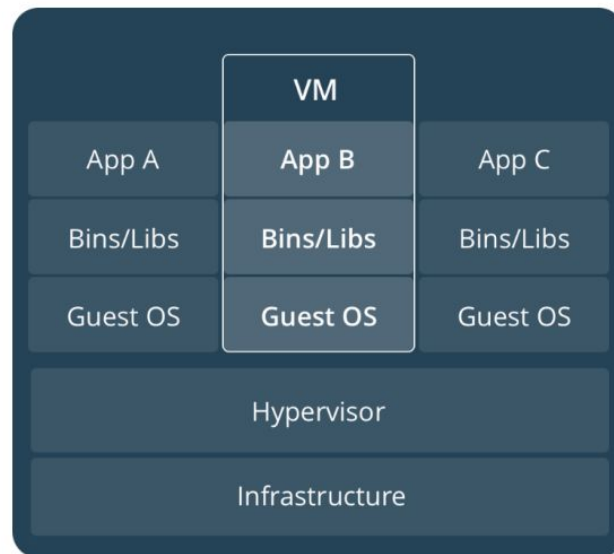
- We have our isolated systems, and after we set up the environment with our colleagues' machine, we expect to get identical results, right? Unfortunately, it is not always the case. Why? Most likely because we run on a different OS.
- Even though using virtual environments, we isolate our computations, we might need to use the same operating system that requires running "like if" we are in different machines.
- How can we run the same experiment? **Virtual Machines!**



# Virtual Machine (VM)

- Virtual machines have their **own virtual hardware**: CPUs, memory, hard drives, etc.
- You need a **hypervisor** that manages different virtual machines on server
- Operating system is called the "**host**" while those running in a virtual machine are called "**guest**"
- You can install a completely different operating system on this virtual machine

(Install a Windows VM on your mac [here](#))



VMs, let's the nostalgics relive  
some of the glory years





# Virtual Machine (VM)

## Pros

- **Isolation:** it works like a separate computer system; it is like running a computer within a computer.
- **Secure:** the software inside the virtual machine cannot affect the actual computer.
- **Costs:** buy one machine and run multiple operating systems.

## Cons

- Uses hardware in your local machine (cannot run more than two on an average laptop)
- Takes time to boot-up
- There is overhead associated with virtual machines

# Where can you use VMs?

- VMs are often use locally, to access a separate OS
- You can also create and use VMs in the **Cloud** !
  - Select specific hardware (e.g. **GPUs**)
  - Pay for what you use
  - Collaborate with your team
  - Isolated and secured environment
    - Often you cannot download customer data on your own machine



Amazon  
**EC2**

[Try it yourself](#)

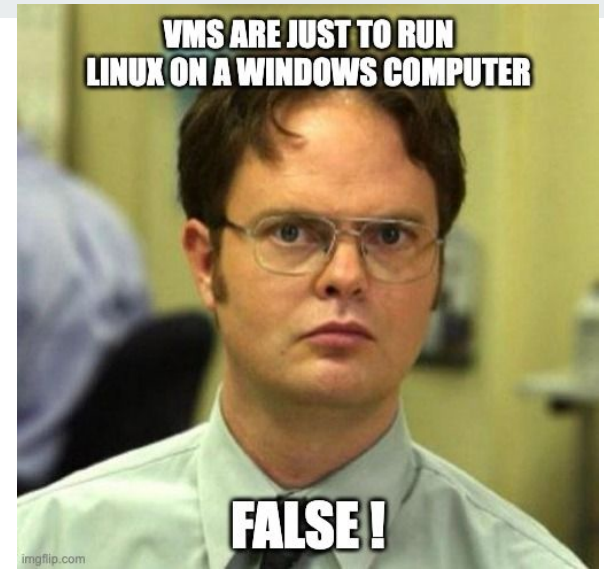


[Try it yourself](#)



Google  
**Compute  
Engine**

[Try it yourself](#)



# Containerisation



# Containers

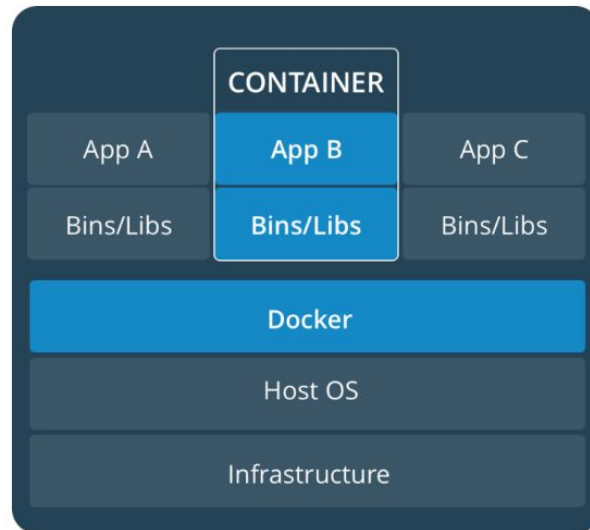


**Containers** encapsulate an application as a **single executable package** that contains all the information to **run it on any hardware**:

- Application code
- Configuration files
- Libraries
- Dependencies

**Abstracts** the application from its **host operating system**.

Containers can be easily transported from a desktop computer to a virtual machine (VM) or from a Linux to a Windows operating system, and they will run consistently on virtualized infrastructures or on traditional “bare metal” servers, either on-premise or in the cloud.



# Docker



Docker is a platform designed to make it easier to create and manage containers.

It is essentially the most popular platform to do so, even though there are alternatives:

- Podman
- rkt
- LXC (Linux Containers)
- containerd
- CRI-o

# What is the difference between an image and container?

**Docker Image** is a template aka blueprint to create a running **Docker container**.

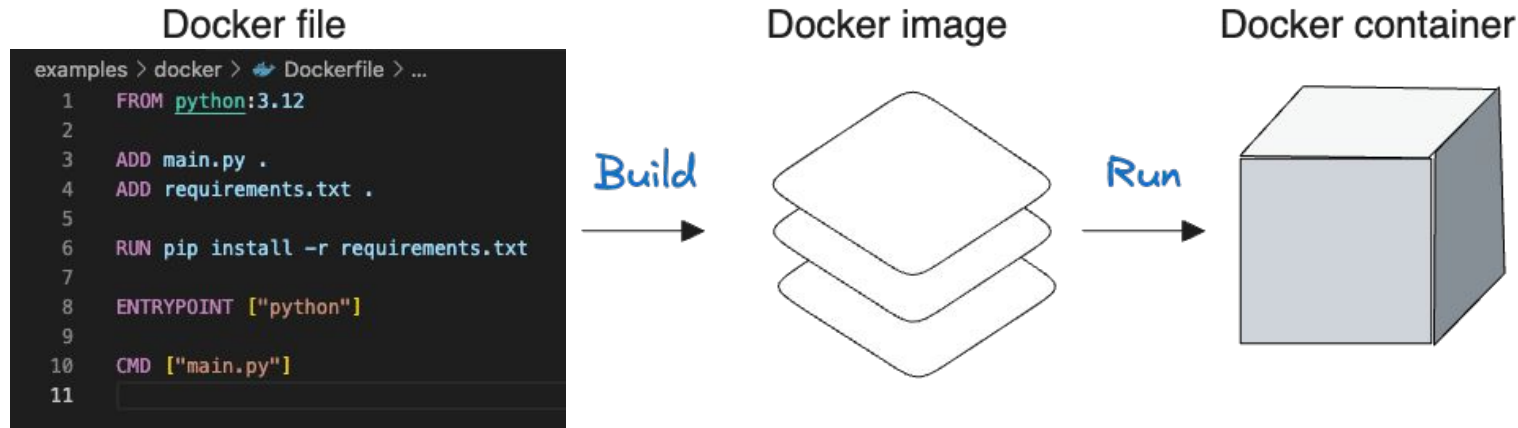
- Docker uses the information available in the Image to create (run) a container.

Image is like a **recipe**, container is like a **dish**.

You can think of an image as a **class** and a container is an **instance** of that class.

# What is the difference between an image and container?

We use the **Dockerfile**, a simple text file, to configure and build the Docker Image. We run the Docker Image to get Docker Container.





# Looking inside a Dockerfile

```
examples > docker > Dockerfile > ...  
1 FROM python:3.12  
2  
3 ADD main.py .  
4 ADD requirements.txt .  
5  
6 RUN pip install -r requirements.txt  
7  
8 ENTRYPOINT ["python"]  
9  
10 CMD ["main.py"]  
11
```

**FROM:** This instruction in the Dockerfile tells the daemon, which base image to use while creating our new Docker image. Here we use a standard Python installation image (already has python installed).

**ADD:** Add source files to the into the container's base folder (you can also add everything with `ADD . .` ).

**RUN:** Instructs the Docker daemon to run the given commands as it is while creating the image. A Dockerfile can have multiple RUN commands, each of these RUN commands create a new layer in the image.

**ENTRYPOINT:** Used when you would like your container to run the same executable every time. Usually, ENTRYPOINT is used to specify the binary and CMD to provide parameters.

**CMD:** The CMD sets default command and/or parameters when a docker container runs. CMD can be overwritten from the command line via the docker run command.

# Popular base docker images

- Tensorflow
- Pytorch
- Python
- Ubuntu
- Alpine
- Nginx
- PostgreSQL
- Redis
- MongoDB

# Multiple containers from same image


*How can you run multiple containers from the same image? Wouldn't they all be identical?*

Yes, you could think of an image as calling a **class**. You can build an image and run it with different parameters using the **CMD** and therefore different containers will be different.

So you can run the same **image** with different **parameters** (e.g. python arguments).

# Multiple containers from same image

Python file with multiple arguments

examples > docker >  main.py > ...

```
1  import argparse
2
3  # Create the parser
4  parser = argparse.ArgumentParser(description="Process some integers.")
5
6  # Add arguments
7  parser.add_argument('--arg1', type=str, default='default_value1',
8                      help='A description for arg1')
9  parser.add_argument('--arg2', type=str, default='default_value2',
10                     help='A description for arg2')
11
12 # Parse the arguments
13 args = parser.parse_args()
14
15 print(f"Argument 1: {args.arg1}")
16 print(f"Argument 2: {args.arg2}")
17
```

# Multiple containers from same image

Dockerfile with multiple CMD options

```
examples > docker > Dockerfile > ...  
1 FROM python:3.12  
2  
3 ADD main.py .  
4 ADD requirements.txt .  
5  
6 RUN pip install -r requirements.txt  
7  
8 ENTRYPOINT ["python"]  
9  
10 CMD ["main.py", "--arg1", "value1", "--arg2", "value2"]  
11
```

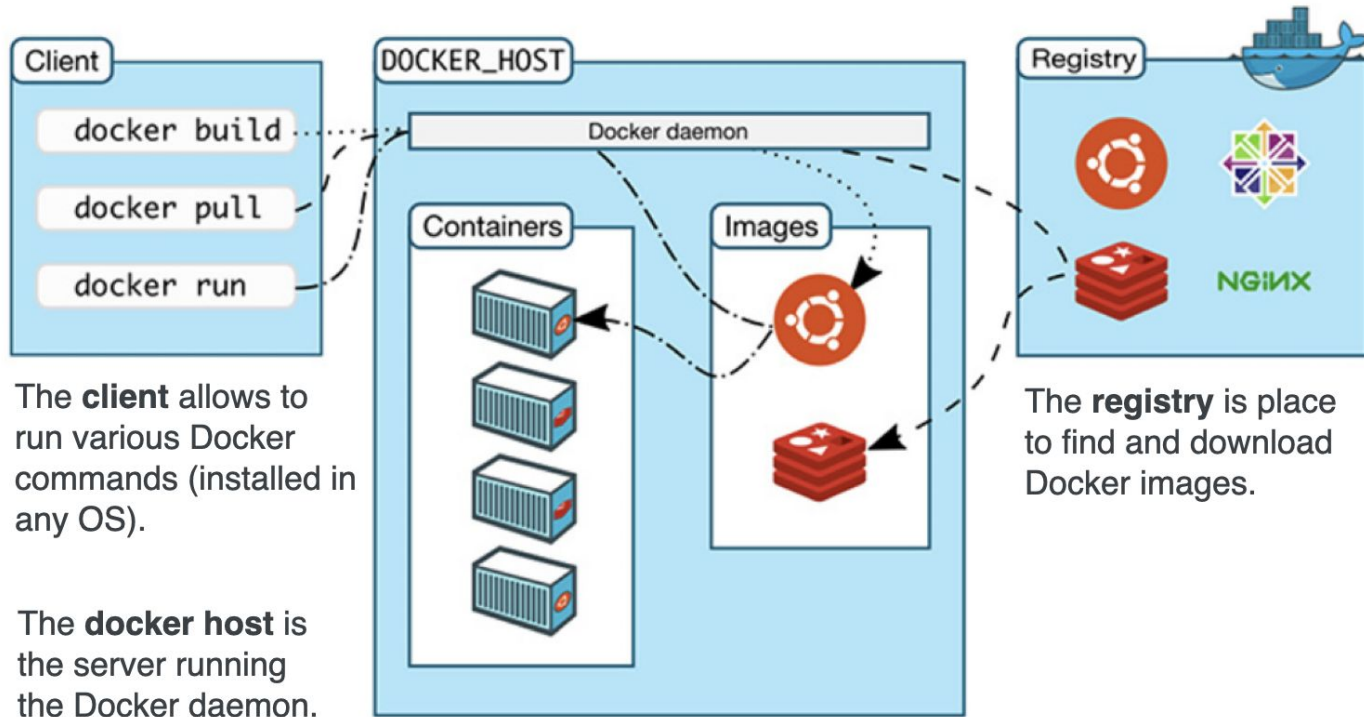
# Multiple containers from same image

Can change python arguments upon docker run

```
• → docker git:(main) x docker run -it python-imagename  
Argument 1: value1  
Argument 2: value2
```

```
• → docker git:(main) x docker run -it python-imagename main.py --arg1 new_value1 --arg2 new_value2  
Argument 1: new_value1  
Argument 2: new_value2
```

# Docker client, host and registry



# Docker registry services

## DOCKER REGISTRY SERVICES



### DOCKER HUB

Docker hub is the official image repository of the docker. Its helps to store, share and distribute the docker image



It is the docker registry owned by Red hat. Its helps to create on premises and cloud repository



### GOOGLE CONTAINER REGISTRY

It is the docker registry created by the google. Its used to setup the private registries

### AMAZON ELASTIC CONTAINER REGISTRY



It is docker registry created by the amazon. This helps the organisation to store and deploy the container in the amazon cloud



# Docker registry services



## GOOGLE CONTAINER REGISTRY

It is the docker registry  
created by the google. Its  
used to setup the private  
registries

Stores your docker image,  
which includes the codes to  
run your application



## Google Cloud Run

Managed service to run your  
application. The hardware  
used is defined here.

# Why should you use containers?



It has the best of the two worlds because it allows:

- to create isolate environment using the preferred operating system
- to run different systems without sharing hardware

The advantage of using containers is that they only virtualize the operating system and do not require dedicated piece of hardware because they share the same kernel of the hosting system.

Containers give the impression of a separate operating system however, since they're sharing the kernel, they are much cheaper than a virtual machine.

# Why should you use containers?



- With container images, we confine the application **code**, its **runtime**, and all its **dependencies** in a pre-defined format.
- With the same image, you can **reproduce** as many containers as you wish.
- A container **orchestrator** is a single controller/management unit that connects multiple nodes together.
  - *To come in “Model Pipeline” lecture!*
- You can create a container on a Windows but install an image of a Linux OS inside that container. The container still works on the Windows machine

# Why should you use containers?



- Containers are **application-centric** methods to deliver high-performing, scalable applications on any infrastructure of your choice.
- Containers are best suited to deliver **microservices** by providing portable, isolated virtual environments for applications to run without interference from other running applications.
- Because they're so **lightweight**, you can have many containers running at once on your system.

# Containers pros & cons

## Pros

- **Portability:** Able to run uniformly and consistently across any platform or cloud.
- **Speed:** Lightweight and only include high level software - fast to modify and iterate on.
- **Efficiency:** OS and infrastructure layer is not contained in the container. Thus, containers are smaller in capacity than a VM and require less start-up time. You can run more containers on the same hardware than VMs.
- **Modularity:** Organise applications into *microservices* that are run independently from each other. Separate development.
- **Fault isolation:** Isolated containerised applications - failure of one container does not affect the continued operation of any other containers. You can identify and correct any technical issues within one container without any downtime in other containers.
- **Ease of management:** Container orchestration platforms (e.g. *Kubernetes*) can ease management tasks such as scaling containerized apps, rolling out new versions of apps, and providing monitoring, logging and debugging, among other functions.

# Containers pros & cons

## Cons

- **Shared host exploits:** Multiple containers often share one hardware. If one container contains an exploit (virus) it could contaminate the entire hardware. Especially as it is common to re-use public pre-made containers.

## ... Also a highly demanded skill !

You need to use it all the time! Therefore highly demanded skill.

*“Coming in at the top of the requirements list, and highlighted in 40% of the total group, was knowledge of container tooling, specifically Docker and Kubernetes. Traditionally, this has been the domain of DevOps, Reliability and Platform Engineers, but it has become a fundamental part of MLOps Engineering. A lack of knowledge in this area puts you at a significant disadvantage to those that do and is likely to be a key development area for those moving into MLOps from ML or Data Engineering backgrounds, who may not have had the opportunity to work on container tooling in production.”*

- Survey on 310 ML Engineers job positions



# Full overview

|             | Virtual environment | Virtual Machine                | Docker |
|-------------|---------------------|--------------------------------|--------|
| Effort      | Easy                | High at beginning,<br>then low | Medium |
| Versatility | Medium              | High                           | High   |
| Portability | Medium              | High                           | High   |



# Lab: Docker

# Directed work 24/02

## To do:

1. Try to redeem your student credits. If you have a problem, we can help you to get them. See Discord for the documentation link
2. Follow the README.md of the directed work.
  - a. Try to do the tutorial (20-30 min).
  - b. Start the homework. This homework should not take you more than 20 minutes so it is possible for you to finish it before the end of the directed work.
  - c. Do not focus on the ML model, so don't take a too large model, too large dataset. It should run on your local machine.
  - d. We'll give you this week the link of the Gradescope page to submit your homework.
  - e. HARD deadline: February 2 2025 23:59.

# Directed work 03/03

## How will it work?

1. You are asked to read again the Docker course AND/OR directed work before Monday morning for it to be fresh in your minds
2. You will receive a homework at the beginning of the directed work by groups (same as for the project).
3. You do it during 45min-1h.
4. 2 groups of students come to the board to present what they have done and why. We discuss this solution together.
5. The other still need to send their homework for the next week.

# Wrap-up

# Project objective for sprint 2

This course focuses on what comes **around** your ML model.

Do **not** spend significant time optimising your data or model.

**No grading** on the **accuracy** of the model itself.

| #   | Week      | Work package   | Requirement |
|-----|-----------|--|-------------|
| 2.1 | W03       | Prepare your data and run an Exploratory Data Analysis.  | Required    |
| 2.2 | W03       | Prepare your Cloud environment. That means creating a Cloud project, granting correct access rights to all members of your group and setting up a billing account.<br><b>Attention:</b> You can have free credits for the Cloud, as explained during the course. | Required    |
| 2.3 | W04       | Train your ML model  | Required    |
| 2.4 | W04       | Evaluate your ML model   | Required    |
| 2.5 | W03 & W04 | Document your data analysis and model performance  | Required    |

# Lecture summary

| Topic                | Concepts   | Relevant for... |      |
|----------------------|--|-----------------|------|
|                      |  | Project         | Exam |
| Cloud infrastructure | <ul style="list-style-type: none"><li>• Cloud providers</li><li>• Cloud services</li></ul>                         | Yes             |      |
| Containerisation     | <ul style="list-style-type: none"><li>• Virtual environments</li><li>• Virtual machines</li><li>• Docker</li></ul> | Yes             | Yes  |
| DW: Docker           |  | Yes             |      |

That's it for today!

