



# HESSIAN

<http://hessian.be>

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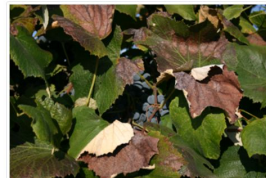
# Introduction

# Introduction (1/2)

- Between 17.2% and 30.3% of global harvests are lost due to plant diseases each year [\[1\]](#).
- **HESSIAN** helps farmers quickly diagnose potential plant diseases before they spread
  - 3 models with different accuracies and costs
  - API and user-friendly online platform

## Model Predictions

Healthy: 15.73%



Sick: 84.27%

## Predictions

🌿 Cassava green mottle

36.56%

🌿 Cassava mosaic disease

30.43%

🌿 Grape black measles

13.32%

😊 Grape healthy

9.77%

😊 Cassava healthy

5.95%

Other

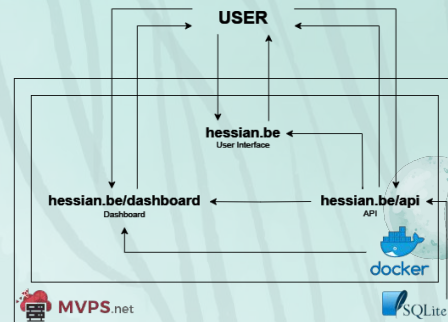
9.97%



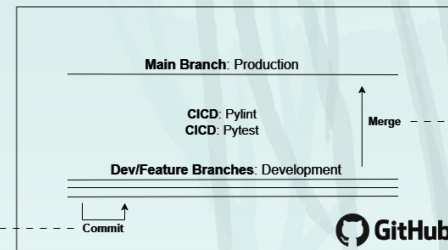
# Introduction (2/2)

- How has **HESSIAN** achieved that?
  - Data collection, analysis, and preprocessing
  - Models selection, training, and evaluation
    - PyTorch, Vertex AI, W&B, GCS
  - Publicly available API
    - Flask, SQLite, Docker
    - Hosted on a VPS
    - <http://hessian.be/api>
  - User interface querying the API
    - HTML, CSS, Javascript
    - <http://hessian.be>
  - Dashboard monitoring the model serving
    - <http://hessian.be/dashboard>
  - CICD
    - Pylint, Pytest, Deployment, Training
  - Gitflow Principles
    - Main branch for production
    - Dev/Feature branches for development

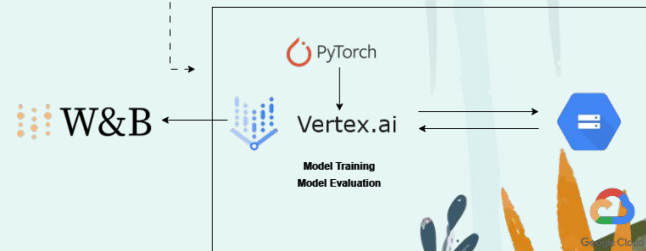
HESSIAN 🌱




CICD: Deployment



CICD: Training



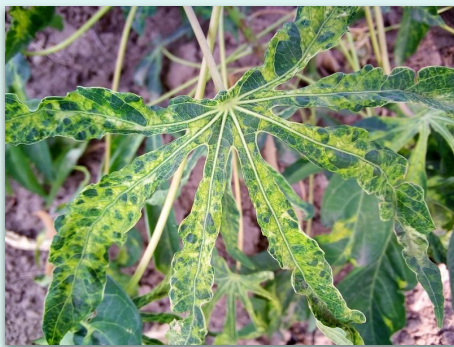


02

# Model Development

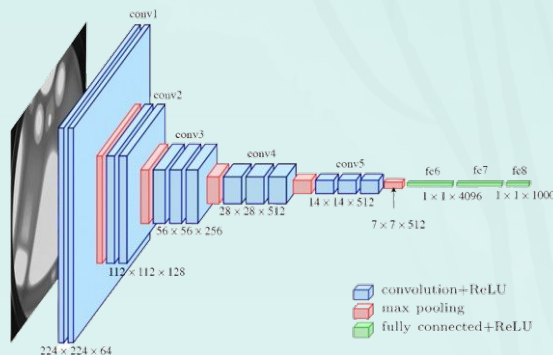
# Model Development: Data (1/3)

- Fixed images dataset
  - 95868 unique images
  - 61 classes
    - 14 healthy
    - 47 ill
  - Imbalance for the Cassava mosaic disease class  $\pm 13000$
- Image transformation
  - Resized to 224x224 format
  - Normalized with the mean and standard deviation of each channel



# Model Development: Models (2/3)

- Inspired by the AlexNet architecture
- Classical PyTorch training pipeline
- 3 different sizes of model
  - Different Accuracies
  - Different Prices

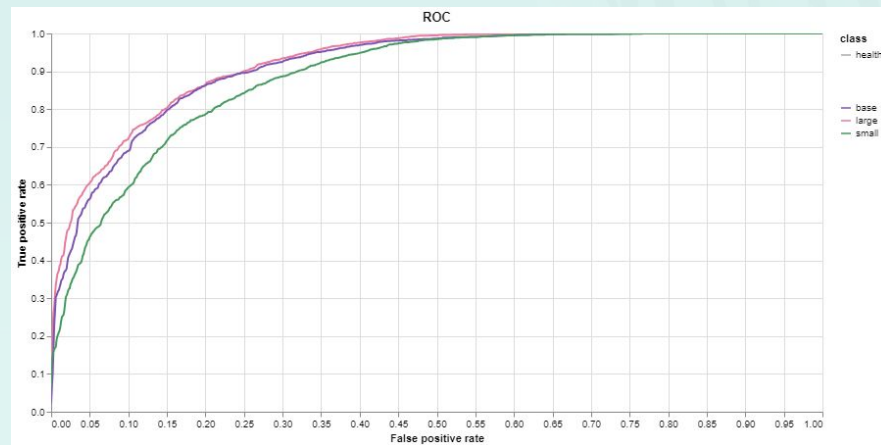


Model Variant	Number of Parameters	Global Accuracy	Binary Accuracy	Costs
Small	134,709	71%	88%	0.02€
Base	526,957	73%	89%	0.04€
Large	2,076,365	76%	90%	0.05€



# Model Development: Monitoring (3/3)

- Training and evaluation monitored with Weights & Biases
  - (Demo [Report](#))





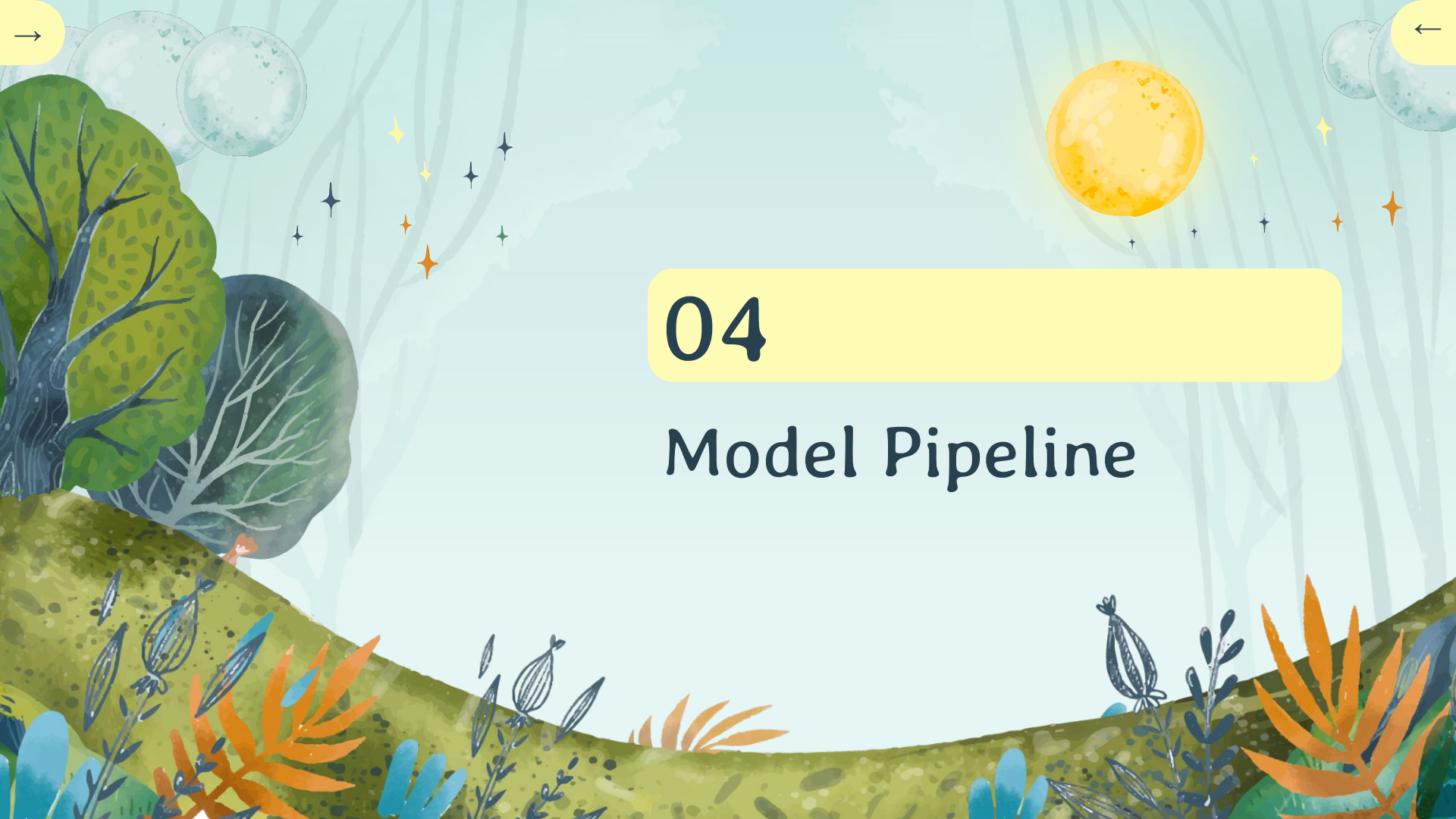
03

# Model Deployment



# Model Deployment: API (1/1)

- Dockerized API based on the Python 3.10.13 image
- API publicly available
  - Queried with HTTP requests
  - Inference requests (image + model size) & Billing details
  - Hosted on a virtual private server (VPS) provided by MVPS
  - Flask to build the web server
  - SQLite to build the database
    - Store requests and billing details
  - <http://hessian.be/api>
  - (Demo Billing Request Postman)
- User interface to query the API in a user-friendly manner
  - <http://hessian.be>
  - (Demo [Image Inference](#))



04


## Model Pipeline





# Model Pipeline: Training (1/1)

- Dockerized training script
  - Usage of Docker volumes to store training data and model weights directly on the host
  - Include W&B monitoring
- Use of Vertex AI to train our models
  - Use of a pre-built PyTorch image
    - Manage dependencies with a custom-made Python source distribution package
  - Use of Google Secret Manager to store sensitive data
  - (Demo [Show Logs](#))
- Use of Google Cloud Storage (GCS)
  - Automatically upload model weights during training
  - (Demo [GSC Organization](#))



05

## Monitoring & CICD

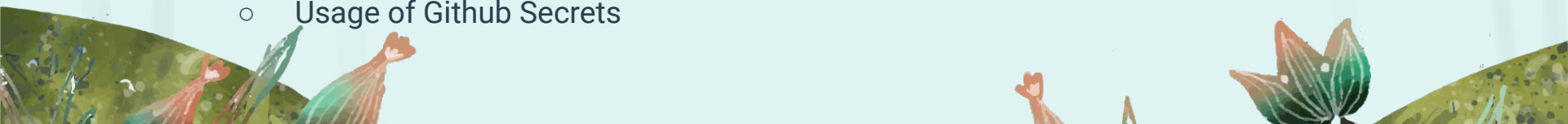


# Monitoring & CI/CD: Dashboard (1/2)

- Dashboard monitoring our application
  - Hardware usage of our server (CPU and RAM)
  - API latency
  - Statistics about API usage
  - <http://hessian.be/dashboard>
- (Demo [dashboard](#))



# Monitoring & CICD: CICD (2/2)

- Using Github Actions
  - **Deployment workflow**
    - Triggered on a merged pull request into main branch
    - Connection to VPS and deploy web server and models ([demo](#))
    - Usage of Github Secrets for VPS connection
  - **Pylint workflow**
    - Triggered by deployment workflow before deploying
    - Checks code compliance with PEP8 guidelines
  - **Pytest workflow**
    - Triggered by deployment workflow once everything is deployed
    - Ensures that the deployed API is working as expected
  - **Training workflow**
    - Triggered when a commit name matches “!train-hessian!”
    - Starts training on Vertex AI ([demo](#))
    - Usage of Github Secrets
- 





# Thanks!

Do you have any questions?

CREDITS: This presentation template was created by [Slidesgo](#), and includes icons by [Flaticon](#) and infographics & images by [Freepik](#)



# Sources

- ◆ [1] Jean B. Ristaino et al. “The persistent threat of emerging plant disease pandemics to global food security”. In: Proceedings of the National Academy of Sciences 118.23 (2021), e2022239118. doi: 10.1073/pnas.2022239118. URL: <https://www.pnas.org/doi/10.1073/pnas.2022239118> (page 1).