Contactless Checking



Contacless checking repository works as client-server connection within microservices running on Kubernetes environment. This repository based on the YOLO-v8, FaceNet, ArcFace serves as checking the people registered and recognize by the system.

The repository using socket connection with Spring Boot web as main hosting APIs to ensure the security connection in each API request. The MYSQL database serves as a relational-database to perform query and storage datasets.

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Developer's Guide

Getting Started

Development Environment

The recommend standard development environment is Ubuntu 18.04 LTS or later. You must install Docker, K8s Cluster Resource or minikube, Helm.

How to use

1. Install docker: docker installation

```
sudo apt-get update
sudo apt-get install ca-certificates curl
sudo install -m 0755 -d /etc/apt/keyrings
sudo curl -fsSL https://download.docker.com/linux/ubuntu/gpg -o
/etc/apt/keyrings/docker.asc
sudo chmod a+r /etc/apt/keyrings/docker.asc

# Add the repository to Apt sources:
echo \
"deb [arch=$(dpkg --print-architecture) signed-by=/etc/apt/keyrings/docker.asc] https://download.docker.com/linux/ubuntu \
```

```
$(. /etc/os-release && echo "$VERSION_CODENAME") stable" | \
sudo tee /etc/apt/sources.list.d/docker.list > /dev/null
sudo apt-get update

sudo apt-get install docker-ce docker-ce-cli containerd.io docker-buildx-
plugin docker-compose-plugin
sudo groupadd docker
sudo usermod -aG docker $USER
newgrp docker
```

2. Install kubectl and helm in test/install 3pp.sh:

```
./install_3pp.sh
```

This will automatic install kubectl and helm mount it to /usr/bin/local and make it global use.

3. Install make

```
sudo apt install make
```

- 4. Building steps are done via make, the builders:
 - o If you don't want to clean the build artifacts, run the following command

```
make build image push
```

If you wan't to clean the build and re-run all the building steps

```
make clean init train build image push
```

The training process will take default hyper-parameters used for YOLOv8, more information, please check: YOLOv8 ultralystics. The hyperparameters can be found in vas.sh.

```
# Hyper parameters
test -n "$TASK_TYPE" || export TASK_TYPE=detect #DEFAULT task=detect is one
of [detect, segment, classify]
test -n "$MODE_TYPE" || export MODE_TYPE=train #DEFAULT mode=train is one of
[train, val, predict, export, track]
test -n "$EPOCHS" || export EPOCHS=50 #DEFAULT EPOCHS=50
test -n "$DEFAULT_MODEL" || export DEFAULT_MODEL="yolov8n.pt" #DEFAULT we
get the pretrained model for training process
test -n "$IMAGE_SIZE" || export IMAGE_SIZE=640
```

5. Using testcon image which integrate the environment for run requirement face_model. Check at test/testcon. In case you don't want to rebuild all necessary library with pip which takes a lot of efforts and time.

In order to using your docker registry. Update DOCKER_REGISTRY in ./vas.sh. Or simply export DOCKER_REGISTRY in your environment.

```
$ export DOCKER_REGISTRY=<your docker-registry>
```

6. Config AWS_ACCESS_KEY_ID and AWS_SECRET_ACCESS_KEY in your env to retrieve all dataset in S3 bucket. Found in IAM roles in AWS.

```
export AWS_ACCESS_KEY_ID=<your-AWS_ACCESS_KEY_ID>
export AWS_SECRET_ACCESS_KEY=<your-AWS_SECRET_ACCESS_KEY>
export AWS_DEFAULT_REGION=<your-AWS_DEFAULT_REGION> | <DEFAULT us-east-1>
```

7. To install helm chart, must build-image-push image to registry before running helm. If could not retrieve the image to pull.

```
$ EXPORT RELEASE=true
$ EXPORT NAME="-n zrdtuan-ns"
$ make package-helm
$ helm $NAME install ck-app build/helm-build/ck-application/ck-application-1.0.0-
5.tgz --set aws.key=$AWS_ACCESS_KEY_ID --set aws.secret=$AWS_SECRET_ACCESS_KEY
```

For TLS, in this lab I config K8s resource in Docker Desktop => Running on WSL. To check the kubernetes IP.

```
$ kubectl get nodes -o wide
```

8. After install helm chart, the container will pull from docker registry to initial the pod running in k8s. Check out the deploy is up and health state.

```
$ kubectl $NAME get all
```

```
NAME
AGE
pod/ck-application-authentication-6d76dd99b7-c4xkr 1/1 Running 0
14m
pod/ck-application-client-56cd64698c-4phls 1/1 Running 0
14m
pod/ck-application-mysql-0 2/2 Running 0
```

14m pod/ck-app 14m	lication-server-84	c4f67c6-ttr	า72	1	/1 Ru	unning	g 0	
NAME		TYPE		CLUSTER	-IP	EXT	ERNAL-IP	
PORT(S)	AGE tion-authenticatio	n LoadBal	lancer	10.106.	76 107	loc:	alhost	
8443/TCP	5s	II LOGGDA	Lancei	10.100.	70.107	1000	J11103 C	
ck-applica 80/TCP	olication-client-http LoadBalancer 10.97.		10.97.2	23.70	loca	alhost		
ck-applica 443/TCP	tion-client-https 5s	LoadBal	lancer	10.108.	149.254	loca	alhost	
ck-applica 3306/TCP	tion-mysql 5s	Cluster	rΙΡ	None		<nor< td=""><td>ne></td><td></td></nor<>	ne>	
ck-application-mysql-read 3306/TCP 5s		Cluster	rIP	10.110.144.98		<none></none>		
ck-applica 5000/TCP	tion-server 5s	LoadBa]	lancer	10.111.	255.161	loca	alhost	
NAME AGE				READY	UP-TO-I	DATE	AVAILA	BLE
	.apps/ck-applicati	on-authenti	ication	1/1	1		1	
deployment 14m	.apps/ck-applicati	on-client		1/1	1		1	
deployment 14m	.apps/ck-applicati	on-server		1/1	1		1	
NAME READY AG	E				DES	IRED	CURREN	Т
READY AGE replicaset.apps/ck-application-authentication-6d7 14m				6d76dd99	b7 1		1	1
replicaset.apps/ck-application-client-56cd64698c 14m				8c	1		1	1
	.apps/ck-applicati	on-server-8	34c4f67c	6	1		1	1
NAME			READY	AGE				
statefulse	t.apps/ck-applicat	ion-mvsal	1/1	14m				

The service data will be manage and stored inside Persistent Volume Claim (PVC), in case we need to reploy the service if crashed, all the data will be preserved, and automatically mounted into pod.

9. Wait a bit untill all pods are running

NAME	READY	STATUS	RESTARTS	AGE
ck-application-authentication-6d76dd99b7-c4xkr	1/1	Running	0	11s
ck-application-client-56cd64698c-4phls	1/1	Running	0	11s
ck-application-mysql-0	2/2	Running	0	11s
ck-application-server-84c4f67c6-ttn72	1/1	Running	0	11s

In the contactless checking system, two server are deploying alongwith one MySQL database for back up and one for primary database, and Web Client. To access into the web for user interface. We need to access into the service.

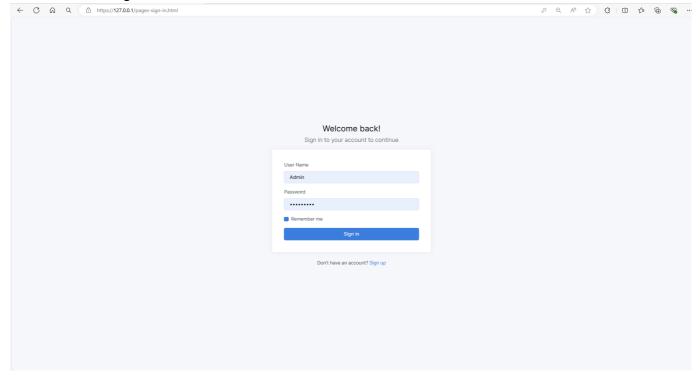
10. Get the service.

```
$ kubectl $NAME get svc
```

This will show all the service to access. Select the Web Client service.

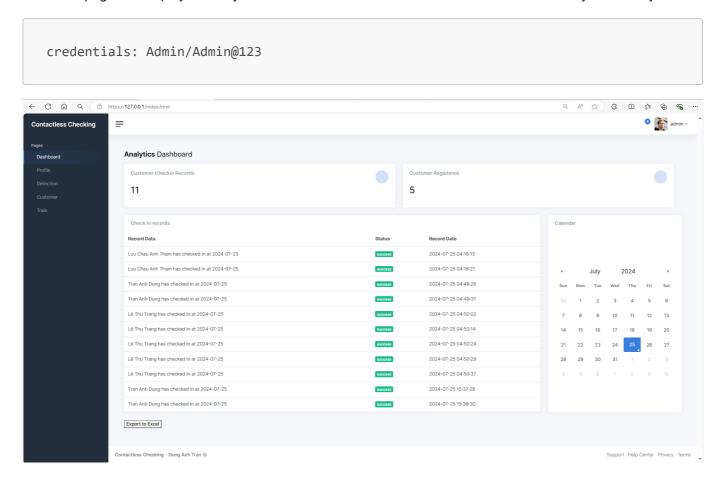
NAME		TYPE	CLUSTER-IP	EXTERNAL-IP
PORT(S)	AGE			
ck-applica	tion-authentication	LoadBalancer	10.106.76.107	localhost
8443/TCP	5s			
ck-application-client-http		LoadBalancer	10.97.223.70	localhost
80/TCP	5s			
ck-applica	tion-client-https	LoadBalancer	10.108.149.254	localhost
443/TCP	5s			
ck-application-mysql		ClusterIP	None	<none></none>
3306/TCP	5s			
ck-application-mysql-read		ClusterIP	10.110.144.98	<none></none>
3306/TCP	5s			
ck-applica	tion-server	LoadBalancer	10.111.255.161	localhost
5000/TCP	5s			

Access https://127.0.0.1 or https://localhost to navigate the Web Client. If access, it will navigate to login page. All the cluster using TLS certificates to authenticate all resources.

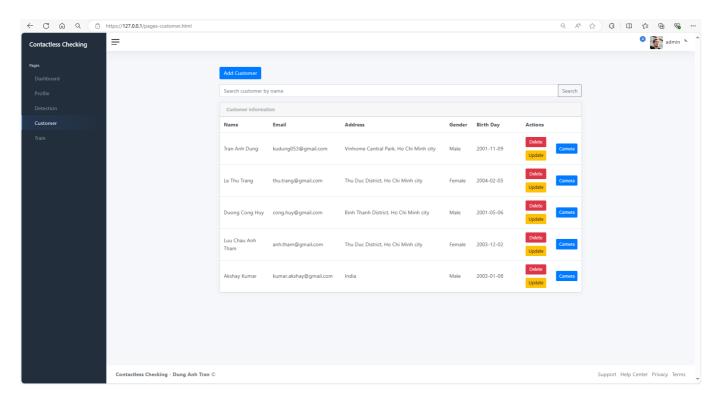


Enter the Username/Password. By default the API Server created default Admin account. Use the credentials to login into pages.

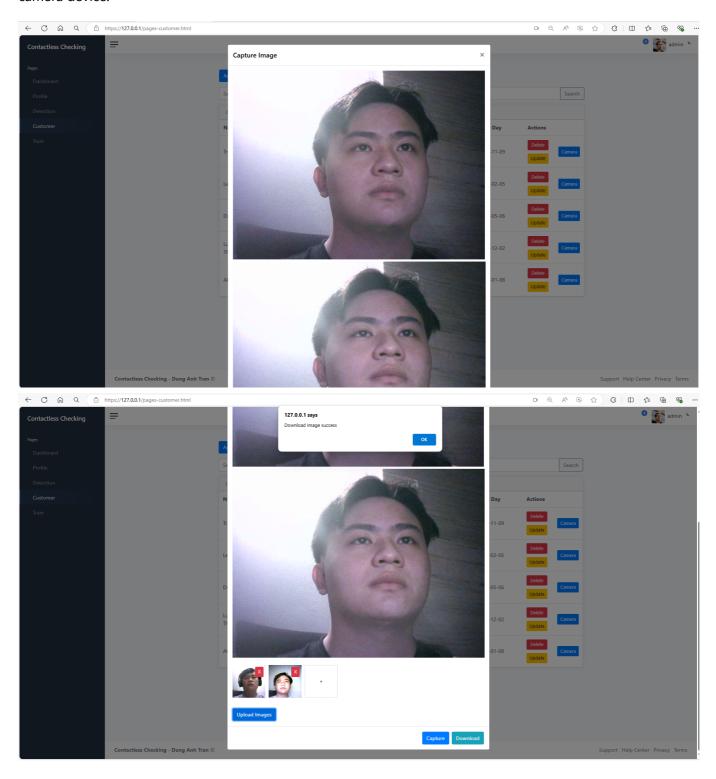
The homepage will display all analytics metrics and record all check-in times, with the ability to filter by date.



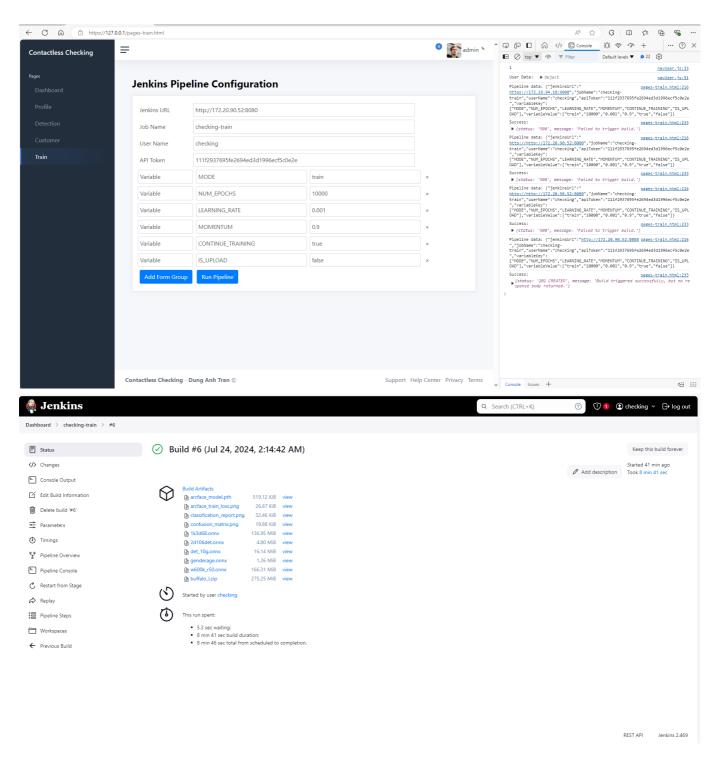
First, create a set of registered customers in the system. Capture all images and send them to the backend, where they will be stored in both a local and a remote S3 bucket to prevent data loss in case of a system crash.



We can edit all customer information and create a new image dataset for the check-in system, which will be sent to train the ArcFaceModel. To obtain image data, we can either upload images or capture them using a camera device.



The collected images will be trained using Jenkins CI. Enter the appropriate parameters and trigger the Jenkins pipeline to train the model with the collected datasets.



The trained model will be used to detect customer images. On the detection page, a snapshot of the customer's image will be sent to the model server for analysis, and the results will be returned to the client.

