VirtualEye - Life Guard for Swimming Pools To Detect Active Drowning Using IBM Cloud

1 INTRODUCTION

1.1 Overview

In modern metropolitan lifestyle, swimming is one of the best activities for stress reduction. Worldwide, drowning results in a higher mortality rate without harming children. The highest global drowning fatality rates are observed to be among children under the age of six. With around 1.2 million incidents each year, these types of deaths rank third among all unexpected deaths worldwide.

1.2 Purpose

The aim of this project is to design a meticulous system which can be implemented along the swimming pools to save human life. An alarm will be issued to call the attention of the lifeguards when the video is being streamed underwater and swimmer position is being examined to

determine the likelihood of drowning. This will help in reducing drowning rates and create a safer environment.

2. LITERATURE SURVEY

2.1 Existing Problem

Beginners, in particular, frequently struggle to breathe underwater, resulting in respiratory issues and, eventually, a drowning disaster. Drowning causes a higher mortality rate worldwide while causing no harm to children. Children under the age of six are found to have the highest global drowning fatality rates. These types of deaths rank third among all unexpected deaths worldwide, with approximately 1.2 million incidents each year.

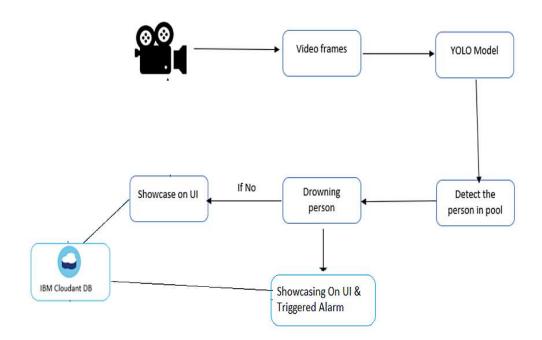
2.2 Proposed Solution

This project builds a model which uses YOLO-based Convolutional Neural Network for object detection. The project makes use of the recent YOLOv3 model for implementation. By studying body movement patterns and connecting cameras to artificial intelligence (AI) systems, the project devises an underwater pool safety system that reduces the risk of drowning. For the scope

of this project, only one camera is used.

3. Theoretical Analysis

3.1 Block Diagram



3.2 Hardware and Software

Software Requirements

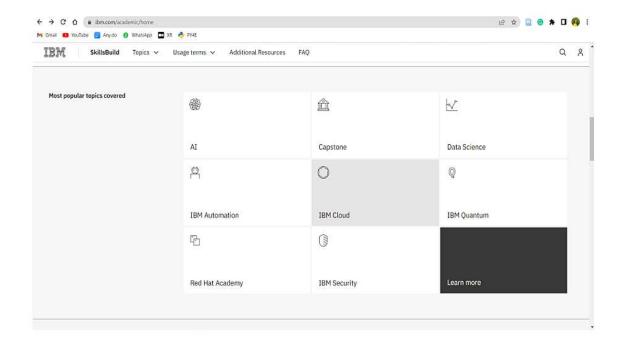
- IBM Cloud
- TensorFlow
- Keras
- Flask

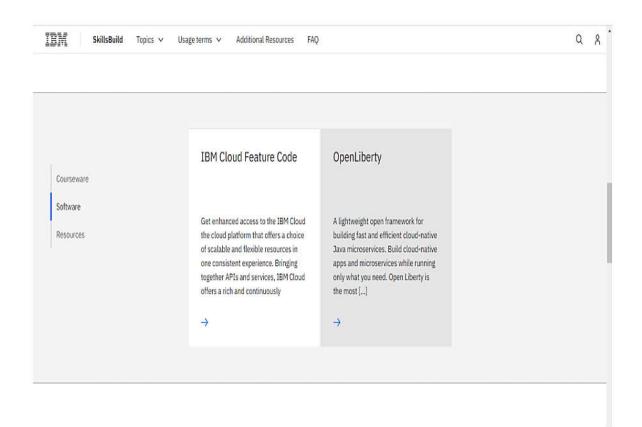
Hardware Requirements

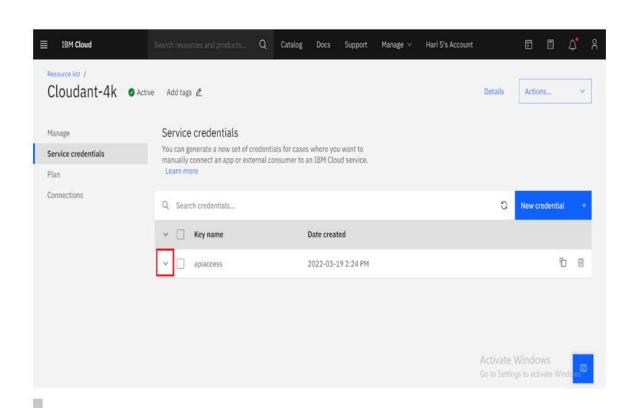
- Processor : Intel Core i3
- Hard Disk Space : minimum 100 GB
- RAM : 4GB
- Display : 14.1" colour monitor (LCD, CRT or LED)
- Clock Speed : 1.67 GHz

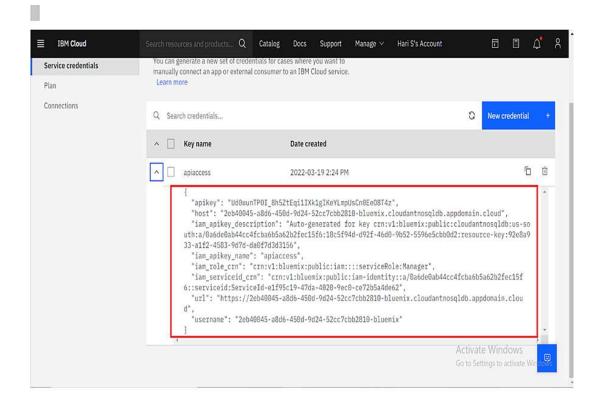
4. IMPLEMENTATION

• Create Database

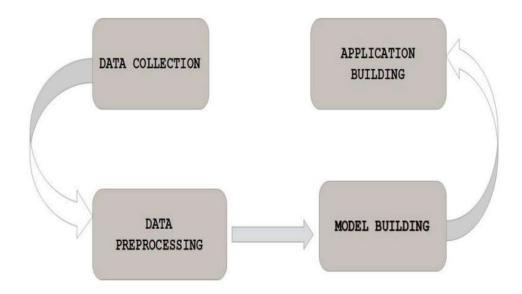




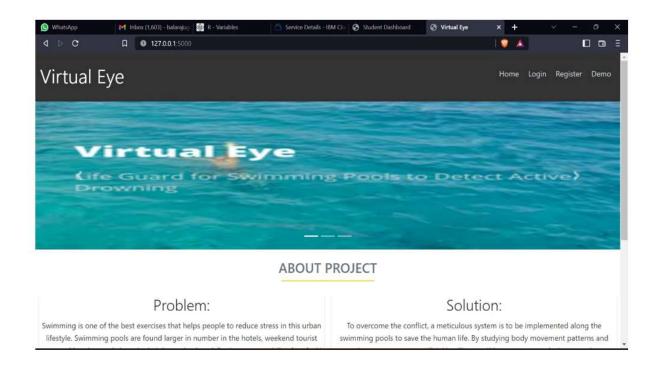


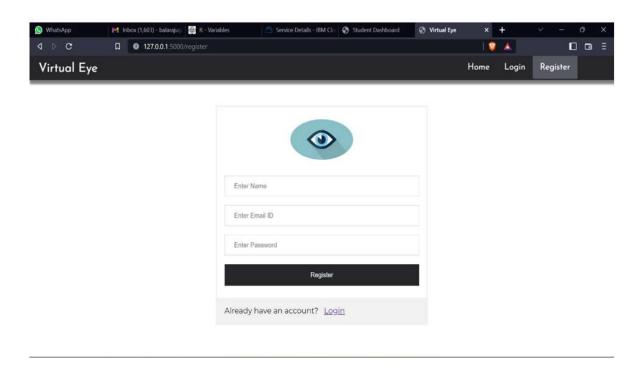


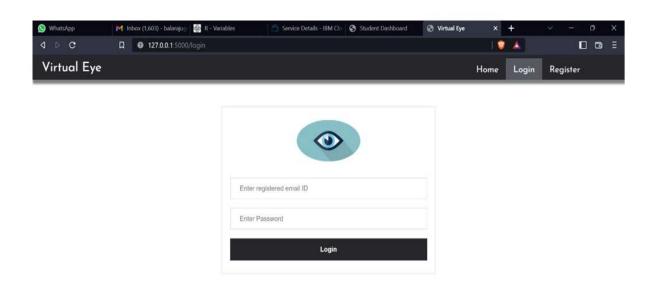
5. FLOWCHART

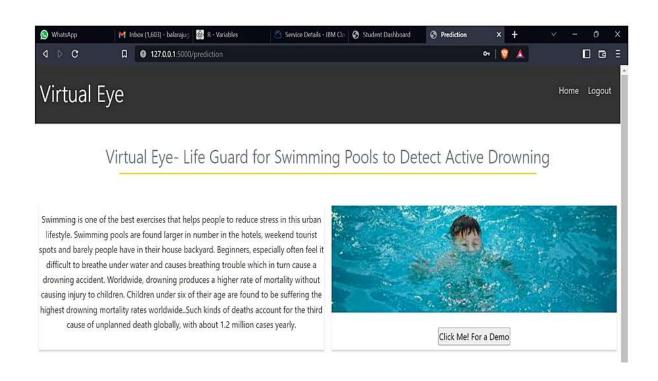


6.RESULT









```
Is he drowning:
                False
2.8001365661621094 s
bbox: [[147, 88, 790, 393], [25, 362, 1265, 540]] centre: [468.5, 240.5] centre0: [469.0, 241.0]
Is he drowning: False
3.686894416809082 s
bbox: [[148, 88, 789, 393], [23, 362, 1265, 539]] centre: [468.5, 240.5] centre0: [468.5, 240.5]
Is he drowning: False
4.606213569641113 s
bbox: [[150, 88, 790, 393], [14, 364, 1272, 539]] centre: [470.0, 240.5] centre0: [468.5, 240.5]
Is he drowning: False
5.5584352016448975 s
bbox: [[149, 88, 790, 394], [9, 365, 1274, 539]] centre: [469.5, 241.0] centre0: [470.0, 240.5]
Is he drowning: False
6.415000915527344 s
bbox: [[149, 88, 787, 393], [8, 382, 1276, 539]] centre: [468.0, 240.5] centre0: [469.5, 241.0]
Is he drowning: False
7.254791021347046 s
bbox: [[148, 89, 790, 393], [5, 383, 1283, 539]] centre: [469.0, 241.0] centre0: [468.0, 240.5]
Is he drowning: False
8.192636966705322 s
bbox: [[148, 89, 791, 393], [5, 381, 1278, 538]] centre: [469.5, 241.0] centre0: [469.0, 241.0]
Is he drowning: False
9.096230745315552 s
bbox: [[147, 89, 791, 393], [2, 381, 1284, 538]] centre: [469.0, 241.0] centre0: [469.5, 241.0]
Is he drowning: False
10.094155311584473 s
bbox: [[148, 89, 788, 393], [-1, 381, 1281, 538]] centre: [468.0, 241.0] centre0: [469.0, 241.0]
Is he drowning: True
127.0.0.1 - - [20/Oct/2022 11:46:54] "POST /result HTTP/1.1" 200 -
```

7. ADVANTAGES AND DISADVANATGES

Advantages:

- The use of deep learning gives accurate results after training the model.
- YOLOv3 model is fast and can process upto 45 frames per second.

Disadvantages:

• YOLO has low recall value and struggles to detect very close objects.

8. APPLICATIONS

- The system can be deployed in homes, hotels, resorts, and swimming pool centres.
- The result is predicted in real-time, thus it can be used in emergency situations.
- The use of deep learning ensures that the prediction is as accurate as possible.

9. CONCLUSION

In this project, we have developed a deep learning system using YOLOv3 model to predict if a person is drowning or not, when underwater. The system is connected to IBM cloud services. The user can access the system through a web application along with an alarm feature system to notify the lifeguards.

10. FUTURE SCOPE

The project can be further extended by deploying multiple cameras underwater to improve accuracy of prediction. The processing speed of the model can be improved to produce the results faster.

11. BIBLIOGRAPHY

- [1] J. Redmon and A. Farhadi, 'YOLOv3: An Incremental Improvement'. arXiv, Apr. 08, 2018. doi: 10.48550/arXiv.1804.02767.
- [2] 'YOLO: Real-Time Object Detection'. https://pjreddie.com/darknet/yolo/ (accessed Oct. 15, 2022).
- [3] H. Aung, A. V. Bobkov, and N. L. Tun, 'Face Detection in Real Time Live Video Using Yolo Algorithm Based on Vgg16 Convolutional Neural Network', in 2021 International Conference on Industrial Engineering, Applications and Manufacturing (ICIEAM), May 2021, pp. 697-702. doi: 10.1109/ICIEAM51226.2021.9446291.

APPENDIX

```
import re
import numpy as np
import numpy as np import os from flask import Flask, app, request, render_template, redirect, url_for from tensorflow.keras import models from tensorflow.keras.models import load_model from tensorflow.keras.preprocessing import image from tensorflow.python.ops.gen_array_ops import concat import cvlib as cv from cvlib.object_detection import draw_bbox import cv2
import cv2
import time
from playsound import playsound
import requests
#Loading the model
from cloudant.client import Cloudant
# Authenticate using an IAM API key client = Cloudant.iam('57f444d5-dfbd-4fc0-b752-dea54005c3cc-bluemix','HTLp9_GkWGDyMR9VHruMMwi_qzZ43qaI3UVR77G0I2GX', connect=True)
# Create a database using an initialized client
my_database = client.create_database('my_database')
app=Flask(__name__)
#default home page or route
@app.route('/')
def index():
     return render_template('index.html')
@app.route('/index.html')
     return render_template("index.html")
#registration page
@app.route('/register')
def register():
     return render_template('register.html')
@app.route('/afterreg', methods=['POST'])
def afterreg():
    x = [x for x in request.form.values()]
      print(x)
      data = {
'_id': x[1], # Setting _id is optional
'name': x[0],
'psw':x[2]
      print (data)
      query = {'_id': {'$eq': data['_id']}}
      docs = my_database.get_query_result(query)
      print (docs)
      print(len(docs.all()))
      if(len(docs.all())==0):
            url = my_database.create_document(data)
#response = requests.get(url)
return render_template('register.html', pred="Registration Successful, please login using your details")
      else:
             return render_template('register.html', pred="You are already a member, please login using your details")
```

```
login page
lapp.route('/login')
lef login():
   return render template('login.html')
!app.route('/afterlogin', methods=['POST'])
lef afterlogin():
   user = request.form['_id']
   passw = request.form['psw']
   print(user,passw)
   query = {'_id': {'$eq': user}}
   docs = my database.get query result(query)
   print (docs)
   print(len(docs.all()))
   if (len (docs.all()) == 0):
       return render_template('login.html', pred="The username is not found.")
       if((user==docs[0][0]['_id'] and passw==docs[0][0]['psw'])):
           return redirect(url for('prediction'))
       else:
           print('Invalid User')
!app.route('/logout')
lef logout():
   return render_template('logout.html')
!app.route('/prediction')
lef prediction():
   return render_template('prediction.html')
!app.route('/result', methods=["GET", "POST"])
```

```
def res():
     webcam = cv2.VideoCapture('drowning.mp4')
     if not webcam.isOpened():
          print("Could not open webcam")
     t0 = time.time() #gives time in seconds after 1970
     #variable docunt stands for how many seconds the person has been standing still for
     centre0 = np.zeros(2)
     isDrowning = False
     #this loop happens approximately every 1 second, so if a person doesn't move,
     #or moves very little for 10seconds, we can say they are drowning
     #loop through frames
     while webcam.isOpened():
          # read frame from webcam
          status, frame = webcam.read()
          #print(frame)
          if not status:
               print("Could not read frame")
               exit()
          exit()
# apply object detection
bbox, label, conf = cv.detect_common_objects(frame)
#simplifying for only 1 person
#print('bbox',bbox)
#print('label',label)
#print('conf',conf)
          #s = (len(bbox), 2)
if(len(bbox)>0):
    bbox0 = bbox[0]
               #centre = np.zeros(s)
centre = [0,0]
               #for i in range(0, len(bbox)):
    #centre[i] =[(bbox[i][0]+bbox[i][2])/2,(bbox[i][1]+bbox[i][3])/2 ]
```

```
centre = [(bbox0[0]+bbox0[2])/2, (bbox0[1]+bbox0[3])/2]
                #make vertical and horizontal movement variables
               hmov = abs(centre[0]-centre0[0])
vmov = abs(centre[1]-centre0[1])
                #there is still need to tweek the threshold
               #this threshold is for checking how much the centre has moved
               x=time.time()
               threshold = 10
               if(hmov>threshold or vmov>threshold):
                     print(x-t0,
                     t0 = time.time()
                     isDrowning = False
               else:
                    print(x-t0, 's')
                     if((time.time() - t0) > 10):
    isDrowning = True
               #print('bounding box: ', bbox, 'label: ' label ,'confidence: ' conf[0], 'centre: ', centre)
#print(bbox, label , conf, centre)
print('bbox: ', bbox, 'centre:', centre, 'centre0:', centre0)
print('Is he drowning: ', isDrowning)
               centre0 = centre
          # draw bounding box over detected objects #print('came here')
          out = draw_bbox(frame, bbox, label, conf,colors=None,write_conf=isDrowning)
          #print('Seconds since last epoch: ', time.time()-t0)
          # display output
          "v2.imshow("Real-time object detection", out)
if(isDrowning == True):
    playsound('alarm.mp3')
                webcam.release()
               cv2.destroyAllWindows()
                "TOOUTH TOHACT COMPTUCE | DUDGITIONT /
          # press "Q" to stop
if cv2.waitKey(1) & 0xFF == ord('q'):
                break
     # release resources
     webcam.release()
     cv2.destroyAllWindows()
     return render template ('prediction.html', prediction="Emergency !!! The Person is drowining")
""" Running our application """
if __name__ == "__main__":
    app.run(debug=False)
```