

Autoencoder Based Gas Detection

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ABSTRACT

The detection of gas emission levels is a crucial problem for ecology and human health. Hyperspectral image analysis offers many advantages over traditional gas detection systems with its detection capability from safe distances. The existing hyperspectral gas detection methods in the thermal range neglect the fact that the captured radiance in the longwave infrared (LWIR) spectrum is better modeled as a mixture of the radiance of background and target gases, we propose a deep learning based hyperspectral gas detection method which combines unmixing and classification.

The proposed method first converts the radiance data to luminance-temperature data. Then, a 3-D convolutional neural network (CNN) and autoencoder-based network, which is specially designed for unmixing, is applied to the resulting data to acquire abundances and endmembers for each pixel. Finally, the detection is achieved by a three-layer fully connected network to detect the target gases at each pixel based on the extracted endmember spectra and abundance values. Furthermore, extension we introduce an Ensemble model that combines three different algorithms: CNN, Bi-directional, and Gated Recurrent Unit (GRU) to improve prediction accuracy.

Imaging spectroscopy has been used by physicists and chemists for more than three decades to identify materials and their compositions. The concept of hyperspectral remote sensing started in the mid-80s and has been widely used by geologists for mapping minerals to this day. The gas leaks in particular in developed countries in the last decade were one of the crucial environmental problems. Some gases are harmful to the environment and contribute to global warming. They present both short-term risks such as explosions and long-term risks such as cancer to workers or people living close to the leaking facility. To minimize these effects, environmental authorities need to monitor chemical and industrial plants to control gas emission levels. Infrared remote sensing technology, which offers many advantages over traditional gas detection systems, is one of the proposed solutions for this aim as such solutions allow monitoring the scene from a safe distance.

2-REQUIREMENTS ANALYSIS

Functional Requirements

2.1.1 Modules:

User:

Input Test Data: Users can input their test values

View Result: It predicts the Gas based on the given test values.

Non - Functional Requirements

Security : Implement robust security measures to protect applicant data.

Scalability : Ability to handle a growing number of applicants.

1-INTRODUCTION

Usability : User-friendly interface that is easy to navigate.

Performance: It works efficiently and Fastly.

Availability : it is available for 24/7 hours.

Portability: it can be used on different operating systems

Computational Resource Requirements

For execution of the project there are some software and hardware requirements.

Programming Language : Python 3.12.3

Front End Frame Work : Flask

Back end Frame Work : Jupyter

Notebook

DataBase : Sqlite3

Front end technologies : HTML,CSS,

JavaScript and

BootStrap4

3-DESIGN

The architecture of this project is designed to ensure seamless interaction between the user interface, data processing, and predictive algorithms. It incorporates modular components for better scalability, maintainability, and performance. Architecture is of two types. They are

Software Architecture

Technical Architecture

Hardware Resources

Operating System : Windows 10

Processor : Intel i5 and above

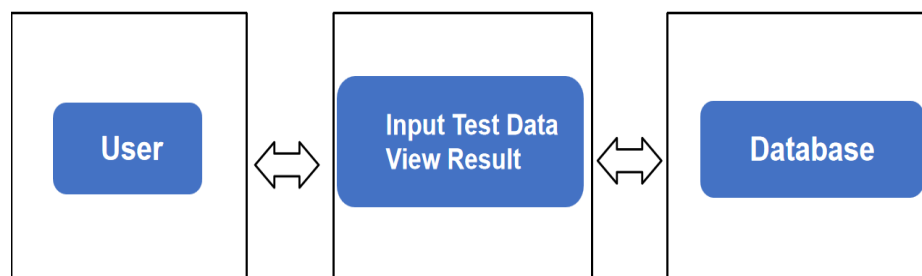
RAM : 4GB

Hard Disk : 500GB

Software Resources

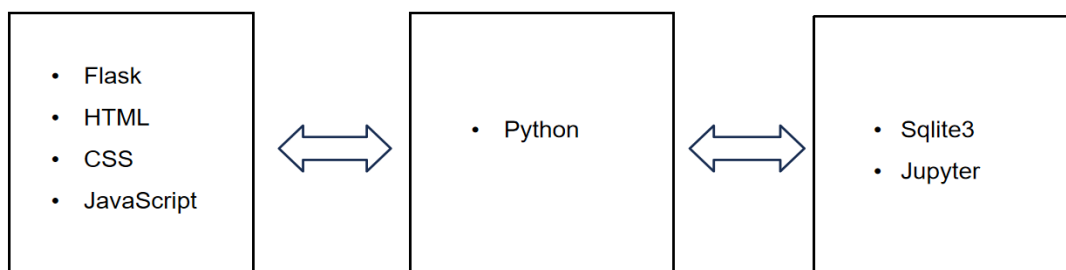
Applicaation : Anaconda

Software Architecture



Software Architecture

Technical Architecture



Data Flow Diagram

Data Flow Diagram (DFD) represents the flow of data within information systems. It provides a graphical representation of the data flow of a system that can be understood by both technical and non-technical users. The models enable software engineers, customers, and users to work together effectively during the analysis and specification of requirements.

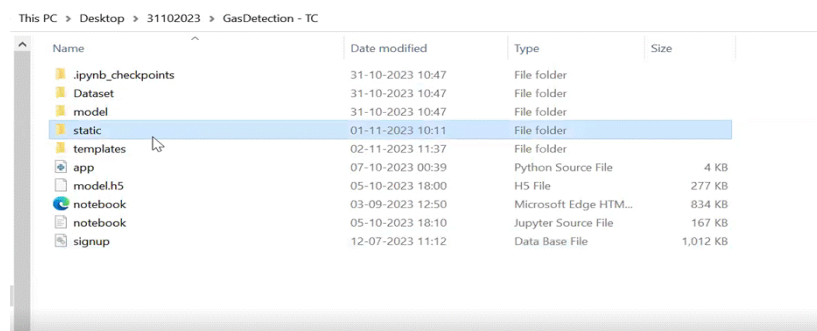
4-IMPLEMENTATION

Python

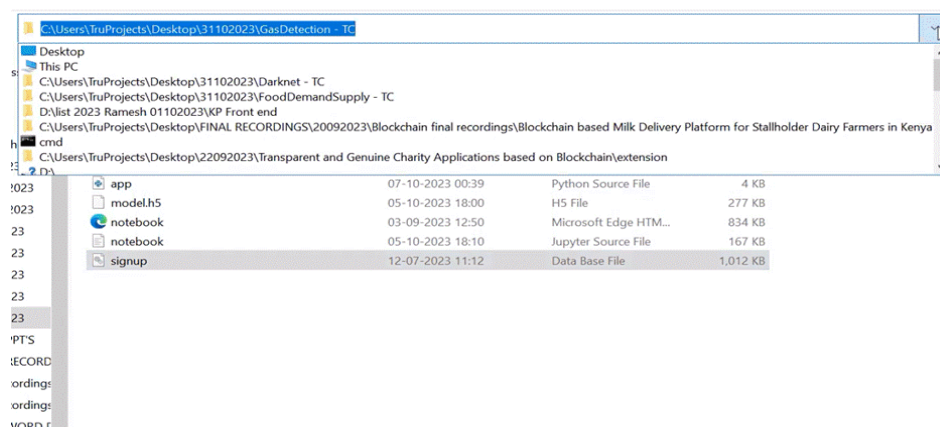
Python is one of the most popular programming languages today, known for its simplicity and extensive features. It was created by Guido van Rossum, and released in 1991. Its clean and straightforward syntax makes it beginner-friendly,

while its powerful libraries and frameworks makes it perfect for developers. It can be used on a server to create web applications. It can be used to handle big data and perform complex mathematics. Python runs on an interpreter system, meaning that code can be executed as soon as it is written. This means that prototyping can be very quick. Python can be treated in a procedural way, an object-oriented way or a functional way. Python relies on indentation, using whitespace, to define scope; such as the scope of loops, functions and classes. Other programming languages often use curly-brackets for this purpose. Python is a versatile, high-level programming language that emphasizes simplicity and readability. It is widely used in various fields, including web development, data science, artificial intelligence, and machine learning, making it ideal for projects.

5-SCREENSHOTS



Screenshot 1 Selecting folder

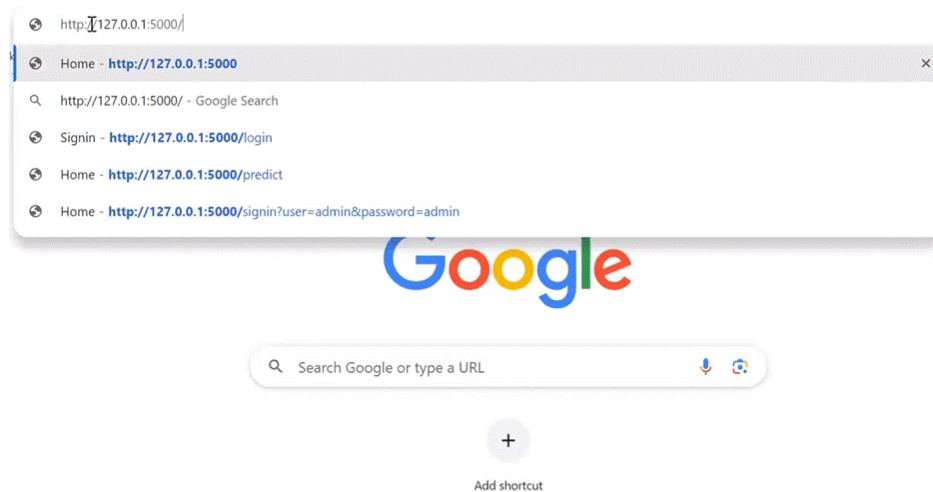


Screenshot 2 Copying folder path

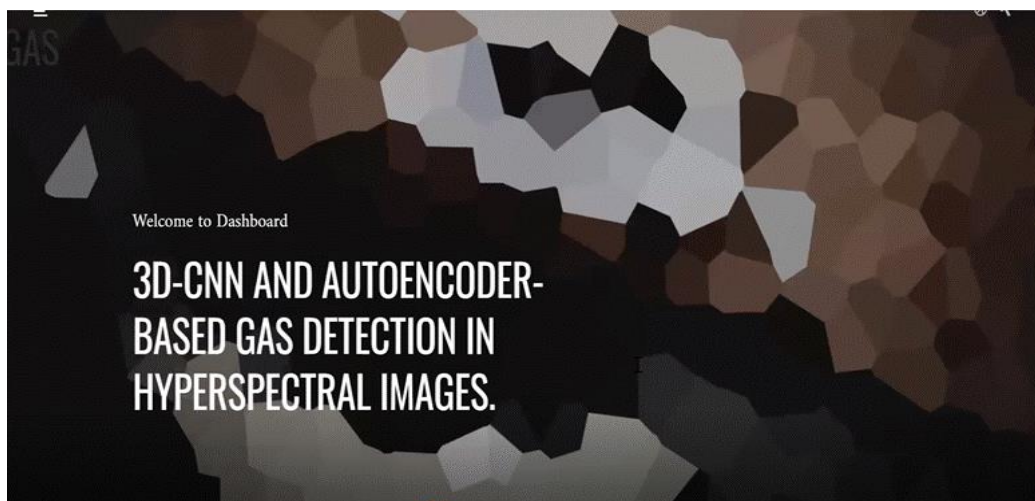
```
Select Anaconda Prompt (Anaconda3) - python app.py

To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.
* Serving Flask app "app" (lazy loading)
* Environment: production
  WARNING: This is a development server. Do not use it in a production deployment.
  Use a production WSGI server instead.
* Debug mode: on
* Restarting with stat
2023-11-03 18:57:18.042437: W tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'cudart64_110.dll'; dlderror: cudart64_110.dll not found
2023-11-03 18:57:18.042803: I tensorflow/stream_executor/cuda/cudart_stub.cc:29] Ignore above cudart dlerror if you do not have a GPU set up on your machine.
2023-11-03 18:57:23.654447: W tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'nvcuda.dll'; dlderror: nvcuda.dll not found
2023-11-03 18:57:23.654661: W tensorflow/stream_executor/cuda/cuda_driver.cc:326] failed call to cuInit: UNKNOWN ERROR (303)
2023-11-03 18:57:23.659353: I tensorflow/stream_executor/cuda/cuda_diagnostics.cc:169] retrieving CUDA diagnostic information for host: DESKTOP-3A1FS42
2023-11-03 18:57:23.659612: I tensorflow/stream_executor/cuda/cuda_diagnostics.cc:176] hostname: DESKTOP-3A1FS42
2023-11-03 18:57:23.660093: I tensorflow/core/platform/cpu_feature_guard.cc:142] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations:
AVX AVX2
To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.
* Debugger is active!
* Debugger PIN: 235-144-891
* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```

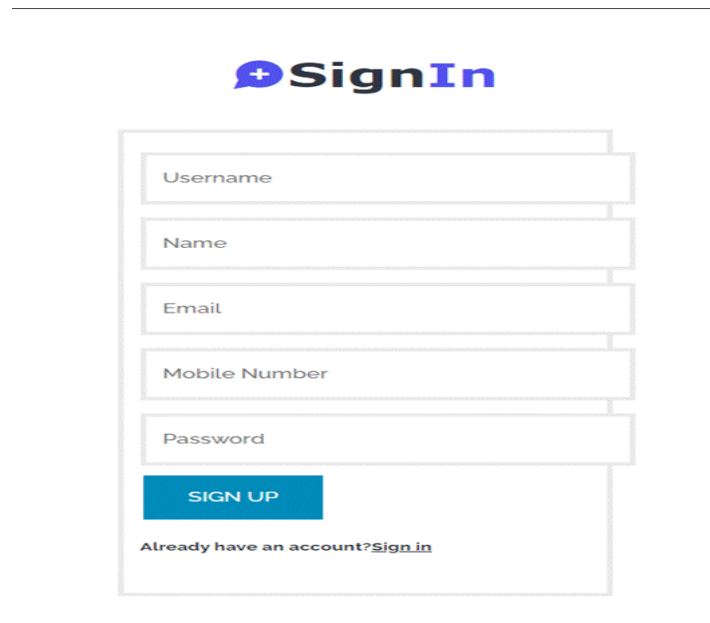
Screenshot 3 Link Generated for webpage



Screenshot 4 Pasting URL in browser



Screenshot 5 Image showing home page of user interface

This screenshot shows a web page titled 'Sign In' with a blue plus icon. The page contains a registration form with the following fields: Username, Name, Email, Mobile Number, and Password. Below the Password field is a blue 'SIGN UP' button. At the bottom of the form, there is a link that says 'Already have an account? [Sign in](#)'.

Screenshot 6 Image showing sign in page

This screenshot shows a web page titled 'Sign In' with a blue plus icon. The page contains a registration form with the following fields: a text field containing 'admin', a password field with four dots, and a blue 'SIGN IN' button. Below the button, there is a link that says 'Register here! [Sign Up](#)'.

Screenshot 7 Image showing sign up of user

X4

X5

X6

X7

X8

X9

X10

X11

PREDICT

Screenshot 8 Image showing Parameters to be entered by user

X1

X2

X3

X4

X5

X9

X10

X11

PREDICT



Screenshot 9 Image showing entered values by user

RESULT: METHANE BASED GAS DETECTED!



Screenshot 10 Results 1

RESULT: SULPHUR BASED GAS DETECTED !



Screenshot 11 Results2

6-CONCLUSION

We have proposed a deep learning-based gas detection method which combines 3D-CNN and autoencoder-based hyperspectral unmixing with neural network-based classification. An ablation study with respect to the possible different combinations for such a system, such as using direct classification methods or using the same structure with other unmixing methods are also performed. In addition, the 3D-CNN and autoencoder-based unmixing has indicated better results than the conventional unmixing for the proposed gas detection framework. The performed study does not require thresholding, unlike the conventional gas detection methods. Finally, the proposed gas detection method achieves better results than state of the art gas detection methods in LWIR range due to its high learning capacity with 3-D convolutional layers. In propose 3DCNN got 88% accuracy, extension model by combining 3 different algorithms

such as CNN + Bidirectional and GRU got 93% accuracy

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