

Streamlined Data Export to CSV

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Abstract- *Fluctuations in weather patterns pose numerous challenges for various sectors, including agriculture, aviation, and infrastructure. Given the unpredictable nature of climatic conditions, accurate weather forecasting has become increasingly crucial. Experts closely monitor technological advancements and evolving atmospheric trends to enhance forecasting accuracy. Weather forecasting plays a vital role in mitigating disasters, facilitating efficient aviation operations, supporting agricultural practices, and meeting diverse business needs. The integration of artificial intelligence (AI) and machine learning (ML) has expanded the scope of weather research by enabling the development of sophisticated models. However, before delving into analysis and model training, meticulous data extraction is paramount.*

In our study, we aimed to identify crucial parameters within radar data sourced from multiple raw files, essential for making precise weather predictions. These parameters include the reflectivity factor, velocity factor, spectrum width, as well as latitude and longitude coordinates. Accurately extracting these radar elements is pivotal for selecting appropriate models, thereby ensuring a system of high accuracy and optimal efficiency.

Keywords- *Weather forecasting, AI and Weather Parameters*

I. INTRODUCTION

Weather forecasting is the process of estimating future atmospheric conditions by utilizing scientific models and methodologies to analyze present weather data. Planning daily activities, agriculture, aviation, emergency preparedness, and other things all depend on it. By giving individuals advance notice of dangerous weather conditions like storms, hurricanes, and heatwaves, accurate predictions save lives by enabling them to take the appropriate safety measures. Forecasts are also necessary for sectors like

transportation, energy, and agriculture to reduce risks and make well-informed decisions.

Regional meteorological centers use radar technology to detect, measure, and track precipitation, which helps in weather forecasting. Meteorologists are able to predict and issue warnings for severe weather phenomena such as storms, hurricanes, and tornadoes in a timely manner because Doppler radar devices provide vital information on wind patterns and storm development.

Radar technology, especially Doppler radar, is commonly used for data extraction utilizing parameters including longitude, latitude, reflectance, spectrum breadth, power and velocity.

Doppler radar is a type of specialized radar that uses the frequency shift of the radar waves reflected off objects to determine their velocity. Doppler radar is essential for detecting the velocity and intensity of precipitation, such as rain, snow, and hail, in weather forecasting. Meteorologists can spot the creation and movement of severe weather phenomena, such as thunderstorms and tornadoes, by analyzing Doppler radar data. This allows for better preparedness and timely warnings, which ultimately improves public safety and reduces damage from hazardous weather occurrences.

Meteorologists can learn more about the composition, intensity, and movement of weather systems by integrating and evaluating various data elements. This data is essential for using monitoring severe weather occurrences and providing reliable weather forecasts.

II. LITERATURE REVIEW

Radar systems are essential for meteorology, providing critical data for weather forecasting. Parameters extracted from radar data, including reflectivity, velocity, power, and geographical coordinates (longitude and latitude), play key roles in understanding atmospheric phenomena. Here's a concise summary of their significance

A. Reflectivity (Z):

Reflectivity measures the intensity of radar signals reflected by precipitation particles. It helps identify precipitation types, estimate rainfall rates, and track storm movement and intensity. Reflectivity is crucial for rainfall estimation, storm tracking, and issuing timely weather warnings.

B. Velocity (Doppler):

Doppler velocity data indicates the speed and direction of precipitation particles relative to the radar site. It provides insights into wind patterns within storms, aiding in the detection and analysis of severe weather phenomena like tornadoes and mesocyclones.

C. Power (P):

Radar power represents the strength of transmitted radar signals, reflecting system sensitivity. Accurate power measurements are essential for calibrating radar systems and ensuring reliable data collection, especially in regions with low signal strength.

D. Longitude and Latitude:

Geographical coordinates specify the location of radar measurements. Longitude indicates east-west position relative to the Prime Meridian, while latitude indicates north-south position relative to the equator. Georeferencing radar data with coordinates enables spatial analysis, storm tracking, and integration with geographical information systems (GIS) for improved weather forecasting accuracy.

These parameters collectively enable meteorologists to analyze weather patterns, detect severe weather events, and issue forecasts and warnings crucial for public safety and disaster preparedness. Integrating radar data with geographic information enhances spatial analysis and visualization, aiding in the interpretation and communication of meteorological information.

III. METHODOLOGY

The web application retrieves radar data elements based on current weather conditions. The reflectivity, power, spectrum width, and velocity values, along with the longitude and latitude obtained from each raw file, are used to estimate the correlation between these radar elements. Longitude and latitude are essential for spatial analysis, which will be used to investigate patterns and correlations between radar data elements across various locations.

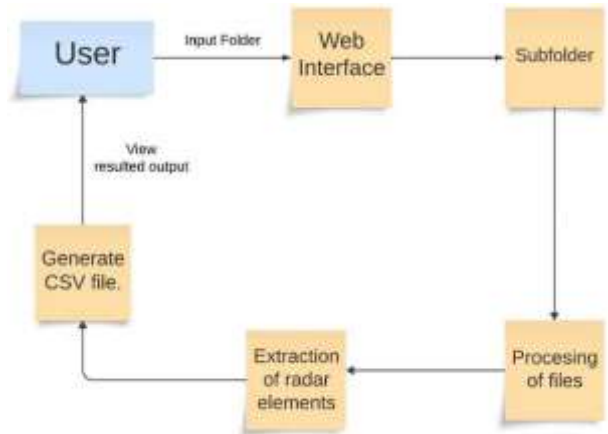


Fig 1. Diagram for Demonstrating Extraction Process

Creating a web application simplifies the process of generating datasets from raw files in batches. Meteorologists have option to download files and use radar data which is present in CSV (Comma-Separated Values) format. This CSV file containing mode values for each parameter will be utilized as input for meteorological analysis. It turned out that the data in the raw file was quite enormous, and the size differed from file to file. This was due to the numbers being dependent on the cloud cover that existed at the time the radar data was collected. For instance, the radar element values for a raw radar file were found in a matrix of size 3600x998. As a result, the data collected for each radar element had to be normalized. NULL values were deleted, superfluous values were filtered out, and the mode value (the most recurring value) of each data file was considered. This analysis considered 289 raw files in a folder.

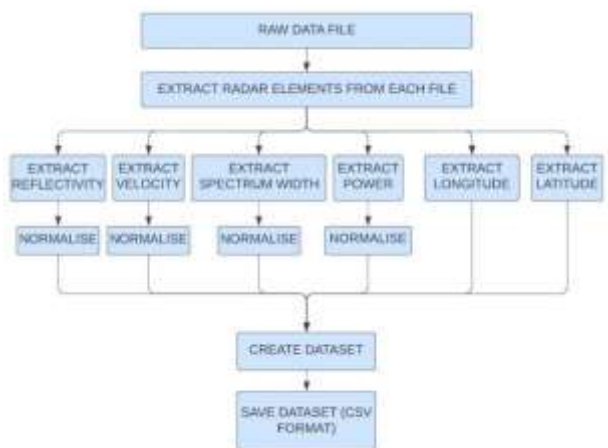


Fig 2. Data Flow Diagram for Parameters Extraction

We used the raw data from a certain day acquired by radar at various intervals. The raw files are a certain kind of file format that contains data in multidimensional array format, which we utilized to extract a specific field using the PY-ART library and then converted to a single CSV to be imported for meteorological analysis.

IV. RESEARCH ANALYSIS

The research project introduces an innovative approach to revolutionize weather forecasting by directly extracting radar parameters from raw data provided by the Indian Meteorological Department (IMD) and structuring them into a user-friendly CSV format. Unlike traditional methods that often involve converting radar data into images, this approach prioritizes simplicity and accessibility. It focuses on extracting crucial parameters like reflectivity factor, velocity factor, spectrum width, and total power, while paying close attention to capturing Min, Max, and MODE values for each parameter. By honing in on MODE values, which represent the most common occurrences within the dataset, the research aims to streamline model training processes and elevate predictive accuracy. The outcomes highlight the effectiveness of this approach in not only improving the efficiency of weather forecasting models but also making them more intuitive and adaptable for users. This breakthrough lays a strong foundation for future advancements in meteorological prediction methodologies, promising enhanced accuracy and reliability in weather forecasts for communities worldwide.

	A	B	C	D	E	F	G
1	time	name	reflectivity	velocity	spectrum	longitude	latitude
2	19:52:49	HYD231128195224.RAWPCMG	31	-1.00787	1.491024	78.47333	17.445
3	13:23:03	HYD231128132224.RAWP9BW	30.5	-1.00787	1.491024	78.47333	17.445
4	10:52:38	HYD231128105223.RAWP82V	30.5	-1.00787	1.491024	78.47333	17.445
5	11:12:41	HYD231128111223.RAWP88M	31	-1.00787	1.491024	78.47333	17.445
6	14:12:32	HYD231128141224.RAWP8SV	30.5	-1.00787	1.118268	78.47333	17.445
7	00:42:55	HYD231128004223.RAWP2XT	30.5	-1.00787	1.491024	78.47333	17.445
8	21:42:28	HYD231128214224.RAWPDK9	30.5	-1.00787	1.118268	78.47333	17.445
9	11:22:43	HYD231128112223.RAWP8BC	31.5	-1.00787	1.491024	78.47333	17.445
10	00:02:48	HYD231128000223.RAWP2KL	30.5	-1.00787	1.491024	78.47333	17.445
11	20:12:52	HYD231128201224.RAWPCU9	30.5	-1.00787	1.491024	78.47333	17.445
12	12:12:52	HYD231128121224.RAWP8SF	32.5	-1.00787	1.491024	78.47333	17.445
13	22:52:40	HYD231128225223.RAWPE5P	30.5	-1.00787	1.118268	78.47333	17.445
14	03:32:44	HYD231128033223.RAWP4C9	31.5	-1.00787	1.118268	78.47333	17.445
15	16:22:53	HYD231128162224.RAWPAW3	30.5	-1.00787	1.118268	78.47333	17.445
16	21:13:03	HYD231128211224.RAWPDB3	30.5	-1.00787	1.118268	78.47333	17.445
17	03:02:39	HYD231128030223.RAWP43U	30.5	-1.00787	1.118268	78.47333	17.445
18	02:02:28	HYD231128020223.RAWP3L2	30.5	-1.00787	1.118268	78.47333	17.445
19	06:32:34	HYD231128063224.RAWP5WG	30.5	-1.00787	1.86378	78.47333	17.445
20	04:22:52	HYD231128042223.RAWP478	30.5	-1.00787	1.491024	78.47333	17.445
21	06:42:36	HYD231128064223.RAWP5Z7	30.5	-1.00787	1.86378	78.47333	17.445
22	10:42:36	HYD231128104224.RAWP804	30.5	-1.00787	1.491024	78.47333	17.445
23	03:12:41	HYD231128031224.RAWP46U	31	-1.00787	1.118268	78.47333	17.445
24	18:22:34	HYD231128182224.RAWPBWL	31	-1.00787	1.491024	78.47333	17.445
25	22:32:36	HYD231128223223.RAWPE0E	30.5	-1.00787	1.118268	78.47333	17.445
26	07:12:40	HYD231128071223.RAWP67R	30.5	-1.00787	1.86378	78.47333	17.445
27	20:42:57	HYD231128204223.RAWPD2F	30.5	-1.00787	1.118268	78.47333	17.445
28	14:52:39	HYD231128145224.RAWFA3S	30.5	-1.00787	1.118268	78.47333	17.445
29	11:52:48	HYD231128115224.RAWP8KJ	30.5	-1.00787	1.118268	78.47333	17.445
30	10:32:35	HYD231128103224.RAWP7XD	31	-1.00787	1.491024	78.47333	17.445
31	05:02:59	HYD231128050224.RAWP549	31	-1.00787	1.491024	78.47333	17.445
32	19:42:47	HYD231128194224.RAWPCL5	30.5	-1.00787	1.491024	78.47333	17.445

Fig 3. Output CSV file

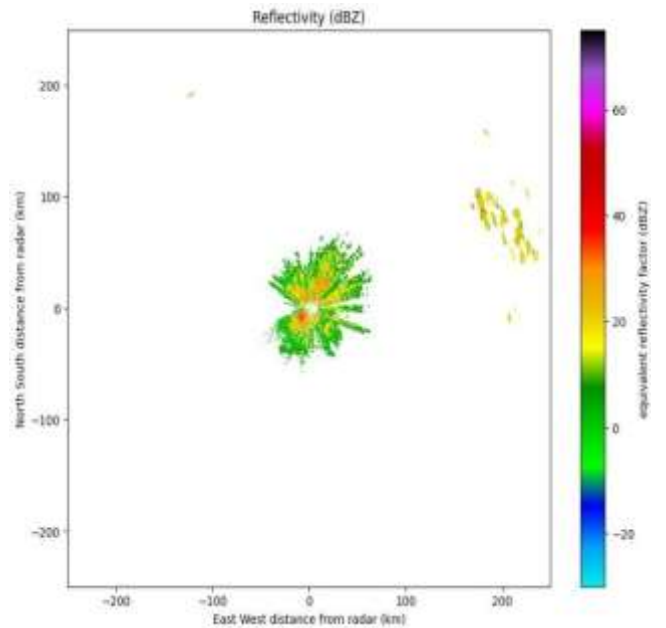


Fig 4. Reflectivity Graph for a radar file

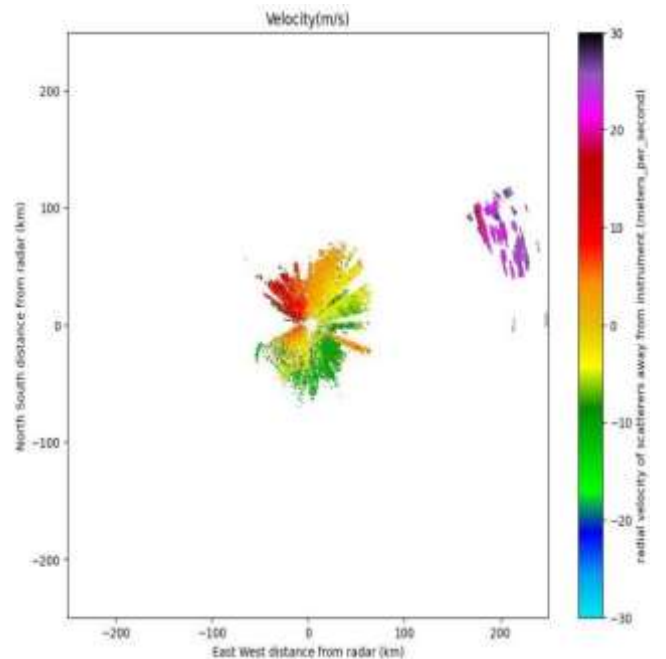


Fig 5. Velocity Graph for a radar file

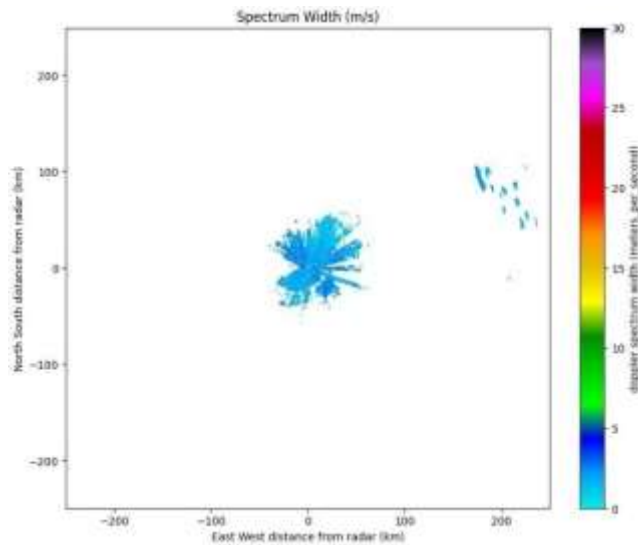


Fig 6. Spectrum Width graph for a radar file

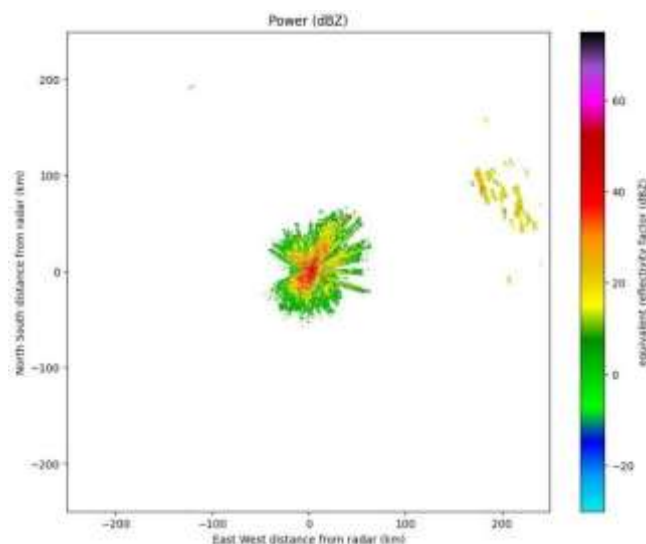


Fig 7. Power Graph for a radar file

V. FINAL REPORT AND ANALYSIS

In summary, there exists a robust correlation between radar power and reflectivity, where higher power yields a more intense reflected signal and heightened sensitivity to targets. To accurately identify and measure targets using radar technology, however, it is imperative to meticulously manage and account for a multitude of complex factors.

Thorough examination of velocity and spectrum width measurements is essential for accurate radar data interpretation. In conclusion, radar velocity and spectrum width are interrelated quantities; however, their correlation is contingent upon various factors and necessitates thorough analysis for proper radar data interpretation.

VI. CONCLUSION

In conclusion, our project successfully tackles the formidable challenge of extracting vital radar data parameters, including reflectivity, velocity, spectrum width, power, longitude, and latitude, from an extensive collection of raw files. By efficiently processing these large datasets into a concise CSV format containing mode values for each parameter, we've achieved a significant milestone in streamlining data preparation for our model. Despite the inherent complexity and size constraints of radar files, our approach ensures that the extracted CSV data remains manageable and conducive to efficient model training.

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