

VGG 16 Architecture

→ One of the most preferred CNN architecture developed by Simonyan and Zisserman by 2014

→ It has 16 convolutional layers

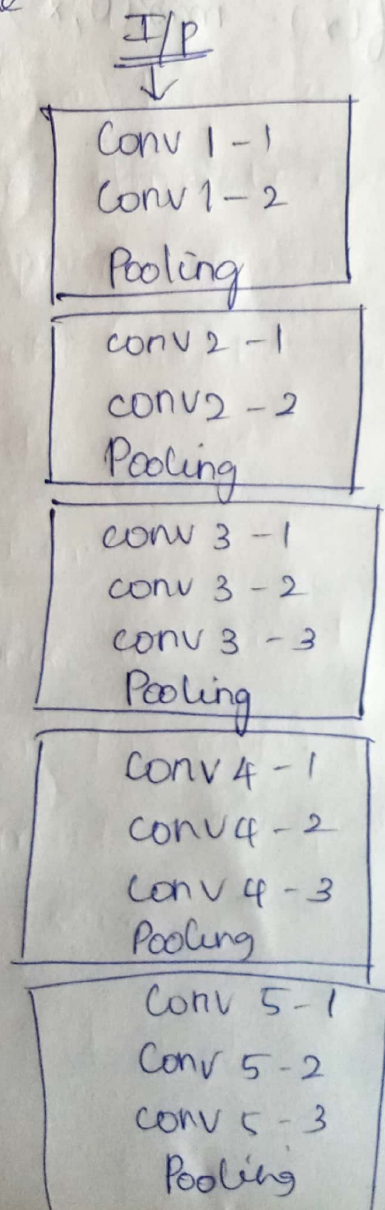
→ VGG 16 is preferred as it has a very uniform architecture.

→ It has huge number of parameters

(i.e.) it has 138 million parameters

which is certainly difficult to handle

Structure



dense
dense
dense



O/p

No	Convolution	O/p dimension	Pooling	O/p dimension
layer 1 1, 2	64 channel 3x3 kernel	224x224x64	Max pool Stride = 2, 2x2 size	112x112x64
layer 3 3, 4	128 channel of 3x3 kernel	112x112x128	Stride = 2 Size 2x2	56x56x128
layer 5, 6, 7	256 channel of 3x3 kernel	56x56x256	Stride = 2 Size = 2x2	28x28x256
layer 8, 9, 10	512 ² channel of 3x3 kernel	28x28x512	Stride = 2 Size 2x2	14x14x512
layer 11, 12, 13	512 channel of 3x3 kernel	14x14x512	Stride = 2 Size 2x2	7x7x512

→ The first layer might be an input layer which could be passed into the function as argument. The function then returns a reference to the final layer in the block, the pooling layer, that could be connected to a flatten

layer and subsequent layers for making a classification prediction.

Tropical Cyclone Intensity Estimation Using VGG

Tropical Cyclone

→ It is one of the costliest disaster.

→ A tropical cyclone is an intense circular storm that originates over warm tropical oceans. It is also called a hurricane or a typhoon.

It is characterized by low atmospheric pressure and heavy rain and its winds exceeds $119 \text{ km (74 miles)}$ per hour.

Formation

→ Once the wind speed increases to 36 km (23 miles) per hour, the storm is classified as a tropical depression.

→ If the circulation continues to intensify and the wind speeds exceed 63 km (39 miles) per hour, then the system is called a tropical storm.

→ Once the maximum wind speeds exceeds $119 \text{ km (74 miles)}$ per hour, the storm is classified as tropical cyclone.

Conditions

1] The temp of surface layer of ocean water must be $26.5^{\circ} (80^{\circ} \text{F})$ or warmer, and this warm layer must be at least $50 \text{ meters (150 feet)}$ deep.

2] A preexisting atmospheric circulation must be located near the surface warm layer.

3] The atmosphere must cool quickly enough with height to support the formation of deep convective clouds.

4] The middle atmosphere must be relatively humid at a height of about 5000 meters (16000 feet) above the surface.

5] The developing system must be at least 500 km (300 miles) away from Equator.

6] The wind speed change slowly with height through the ^{tropics} atmosphere - no more than 10 meters (33 feet) per sec b/w surface and altitude of about 10,000 mtrs (33000 feet).

Research Paper

⇒ First of all, they used 5 kt interval estimation of wind speed with data available from Naval Research

Laboratory. Now it extends by using 1 kt interval, and using real-time

Geo-stationary Operational Environment

Satellite Imagery, performing

extensive evaluation and building

and deploying a production system.

Training dataset

- i) Identify HURDAT2 storm intensity, time, location (latitude, longitude of storm center).
- ii) Create a bounding box around storm using start and end date time of storm.
- iii) Use bounding box and time to download GOES-8, 10, 11, 12, 13, 16.
- iv) Create a padding of ± 5 degree from center of storm on both latitude and longitude.
- v) Match HURDAT2 wind speed to the closest file if there is not an exact match in time.
- vi) Interpolate location information and wind speed (1kt interval).
- vii) Apply random rotation, random shear, zoom on training data to create more training samples.

Using this training dataset, design a deep learning model for objective estimation of tropical cyclone intensity using CNN.

⇒ This model inputs training samples at 5kt speed intervals and outputs a maximum wind speed at 1kt resolution, however, model precision cannot exceed the 5kt resolution of i/p training data.

Performance Matrix

i) Mean Absolute Error = $\frac{1}{n} \sum |x_p - x_t|$

ii) Root mean Squared Error = $\sqrt{\frac{1}{n} \sum (x_p - x_t)^2}$

iii) Bias = $\frac{1}{n} \sum (x_p - x_t)$

iv) Relative Root Mean Squared Error

$$(CRMSE) = \frac{\sqrt{\frac{\sum (x_p - x_t)^2}{n-1}}}{x_p}$$

x_p ⇒ predicted intensity.

x_t ⇒ actual intensity n ⇒ no. of samples

In CNN model, ^{variation of} VGG 16 architecture is used to estimate. This model includes 4 convolutional layers where each layer is followed by max pooling layer and a dense layer is included, which uses linear activation.

Model Hyperparameters :-

- i) Learning rate of $1e-5$
- ii) Batch size - 60
- iii) ReLU - Activation function
- iv) Pooling with overlaps
- v) Adaptive Moment Estimation (Adam) Optimizer

⇒ It includes tracing the CNN's final intensity back to original image to discover which pixel contributed most to the classification by using Class Activation Map (CAM).

- This model is extensively evaluated and systematically transitioned to production by comparing features identified in CAMS to Dvorak Techniques
- CNN is learning the desired cloud features for classification, particularly at higher intensities.