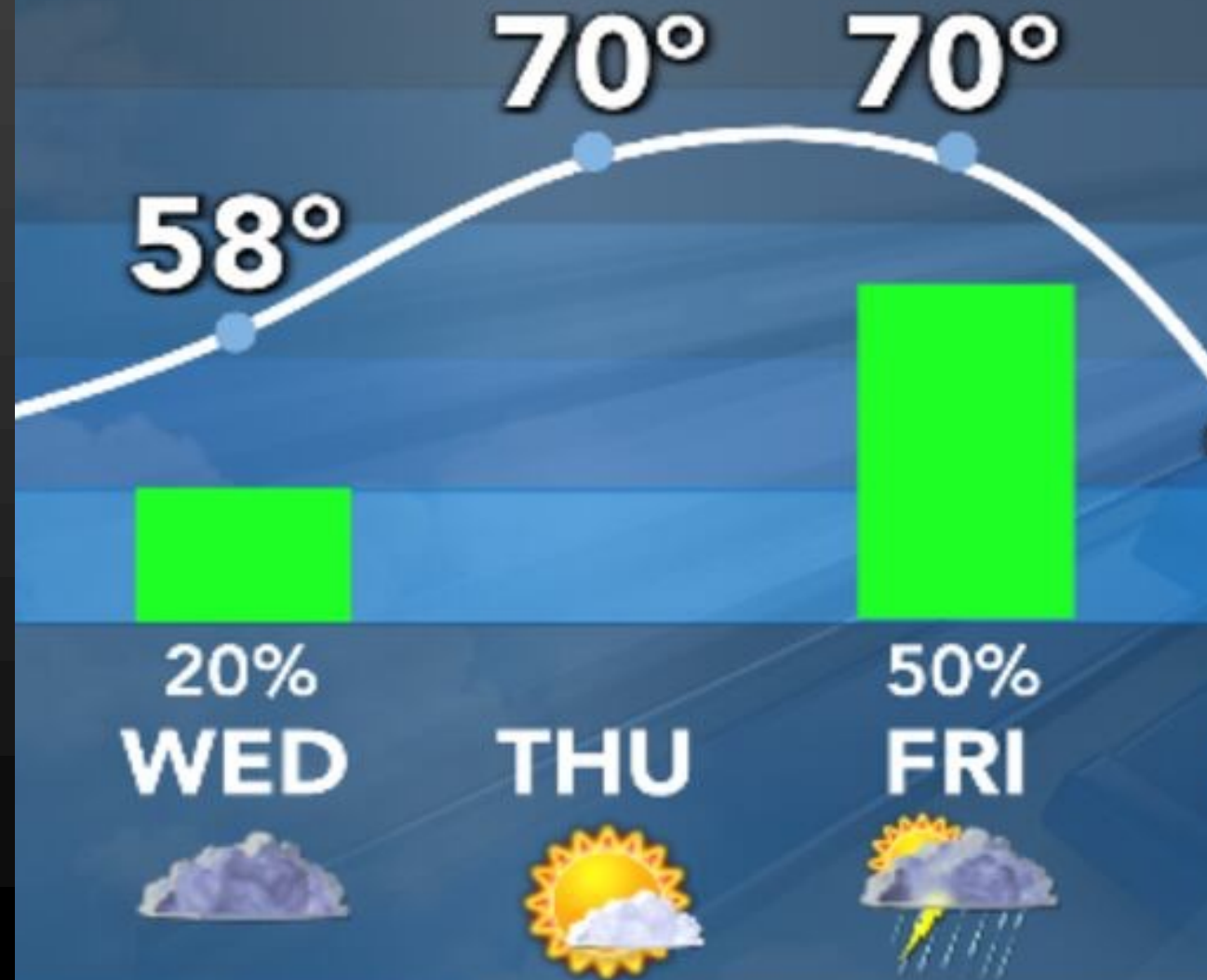


Long-range weather forecasting using Radial Basis Function

A presentation at ICAAN

Cesar Kabue Waweru

CAST TREND



About Me

Cesar Waweru Kabue (CWK)

- print(**“Who am I”**)
 - Problem-solver
 - Cyber security expert
 - Computer Scientist and Researcher
 - Application Tester
 - Son, brother and friend





INTRODUCTION

1. **What is weather?**
2. **What is weather forecasting**
3. **Existing models**
4. **What is the problem?**
5. **What is the research gap Vs research objectives?**
6. **Radial Basis Function Network**
7. **North-Eastern Kenya**
8. **Results**
9. **Demo**
10. **Conclusion**

WEATHER VS PREDICTION

1. The state of the atmosphere of a location/ place.
2. Use of science and technology

High **51°F** low **33°F**

Chance of precipitation **24%**



Typical precip: **0.4in**

Typical Humidity: **71%**

Typical Wind: **13.9mph**

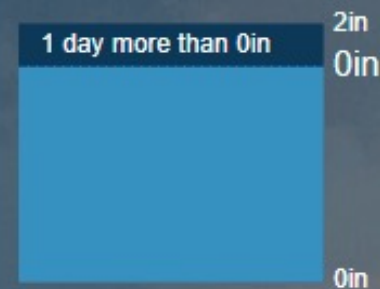
Temperature

In a typical week



Precipitation

In a typical week



Wind

In a typical week



Humidity

In a typical week



Hourly detail

Morning

Temperature: 41°F
Precip chance: 8%

Afternoon

Temperature: 48°F
Precip chance: 9%

Evening

Temperature: 41°F
Precip chance: 8%

Night

Temperature: 38°F
Precip chance: 7%

WEATHER FORECASTING MODELS

SCIENTIFIC TECHNOLOGICAL MODELS (MACHINE LEARNING/DEEP LEARNING)

1. Artificial Neural Network Models
 1. Radial Basis Function Network (RBFN)
 2. Back Propagation Network (BPN)
 3. Multi-layer Perceptron Network (MLPN)
 4. Elmans Recurrent Neural Network (ERNN)
2. Support Vector Machines (SVM)
3. K Nearest Neighbour Model (KNN)
4. Genetics Algorithm (GA)
5. Generic Programming (GP)
6. Random Forest (RF)
7. Hybrid Models

NUMERICAL WEATHER PREDICTION MODELS

1. Statistical models
2. Mathematical models
3. Satellite-based Models.



PROBLEMS IN WEATHER FORECASTING

1. Computational Complexity- Solution to Weather Differential Equations.
2. Level of accuracy of the weather prediction model.
3. Mapping of weather model with the problem of interest.
4. Chaotic/ Unpredictable nature of weather



RESEARCH GAP VS OBJECTIVES

RESEARCH GAP

1. Only short-range (Mins to hours) weather forecasting model.
2. Single-Parametric Prediction Approach.
3. Few weather prediction research conducted in the African location.

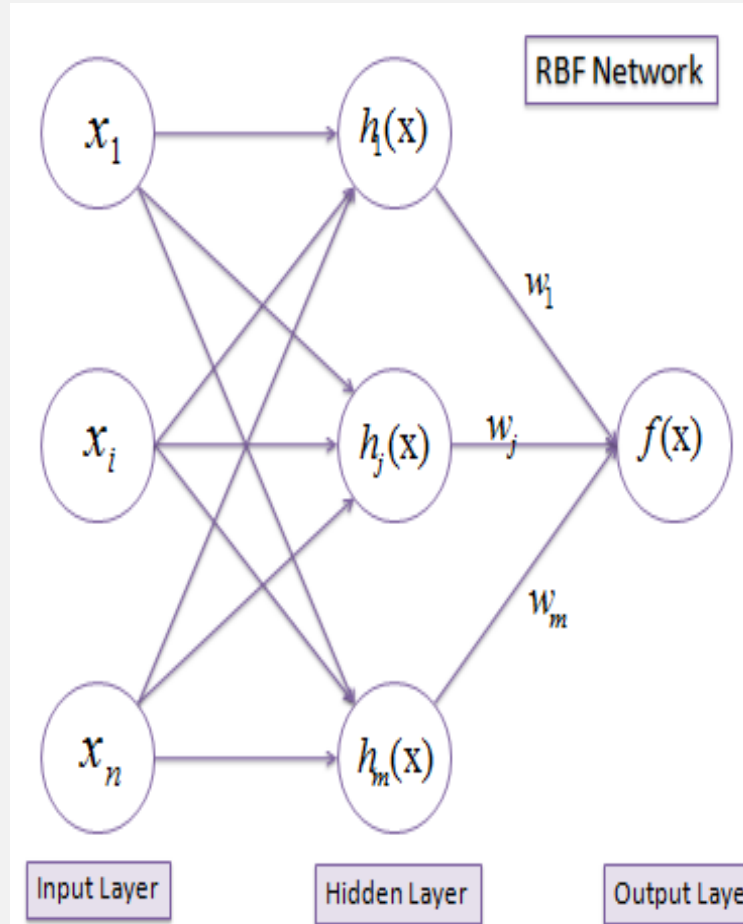
RESEARCH OBJECTIVES

1. Long-range (days, weeks, months) weather forecasting model.
2. Multi-Parametric Prediction Approach.
3. Case Study of Kenya-North Eastern, Marsabit



RADIAL BASIS FUNCTION NEURAL NETWORK

It is a customized instance of a neural network that uses the radial basis function as the activation function for the weights of the input variables.



$$f(x) = \sum_{j=1}^m w_j h_j(x)$$

$$h(x) = \exp\left(-\frac{(x-c)^2}{r^2}\right)$$

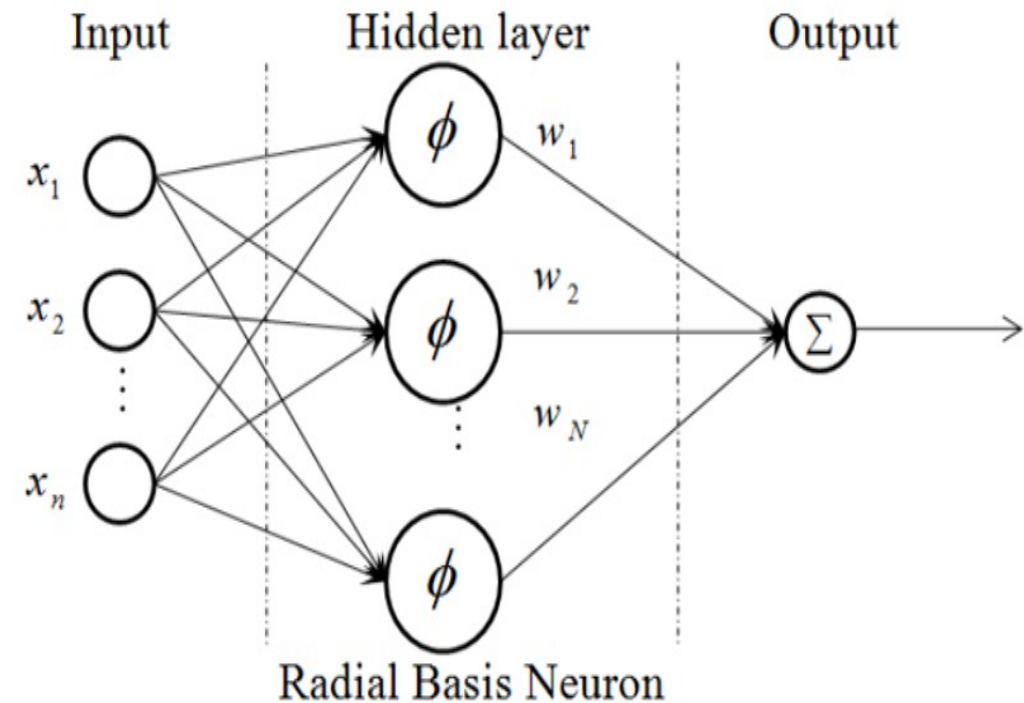
RBFN FOR LONG-RANGE WEATHER PREDICTION

INPUT VARIABLES

X VARIABLE	SI UNIT
Temperature	(°C)
Humidity	$(\frac{gm}{m_3})$
Windspeed	(m/s)
Sea Level	(p)

OUTPUT VARIABLE

Y VARIABLE	SI UNIT
Rainfall	(mm)



LOWER LEVEL IMPLEMENTATION

LAYER TYPE	VARIABLE	UNITS
Input Layer	1. Temperature 2. Humidity 3. Windspeed 4. Sea Level	5
Hidden Layer (RBF Neuron)	Gaussian function	10 50 100
Output Layer	Rainfall	1



KENYA-NORTH EASTERN KENYA

The East African region, according to reports is one of the fastest-growing economies in Africa. Inside it, Kenya is ranked as the fastest-growing economy with an annual average growth of 5% between 2010 and 2018, with a Gross Domestic Product (GDP) of USD 95 Billion. Its economy is described as diverse and dynamic by USAID. Furthermore, it serves as an entry point for the larger East Africa. (USAID, RSM Consulting)

In Kenya's GDP, approximately 33% of this is agriculture, greatly affected by the level of accuracy of weather prediction.

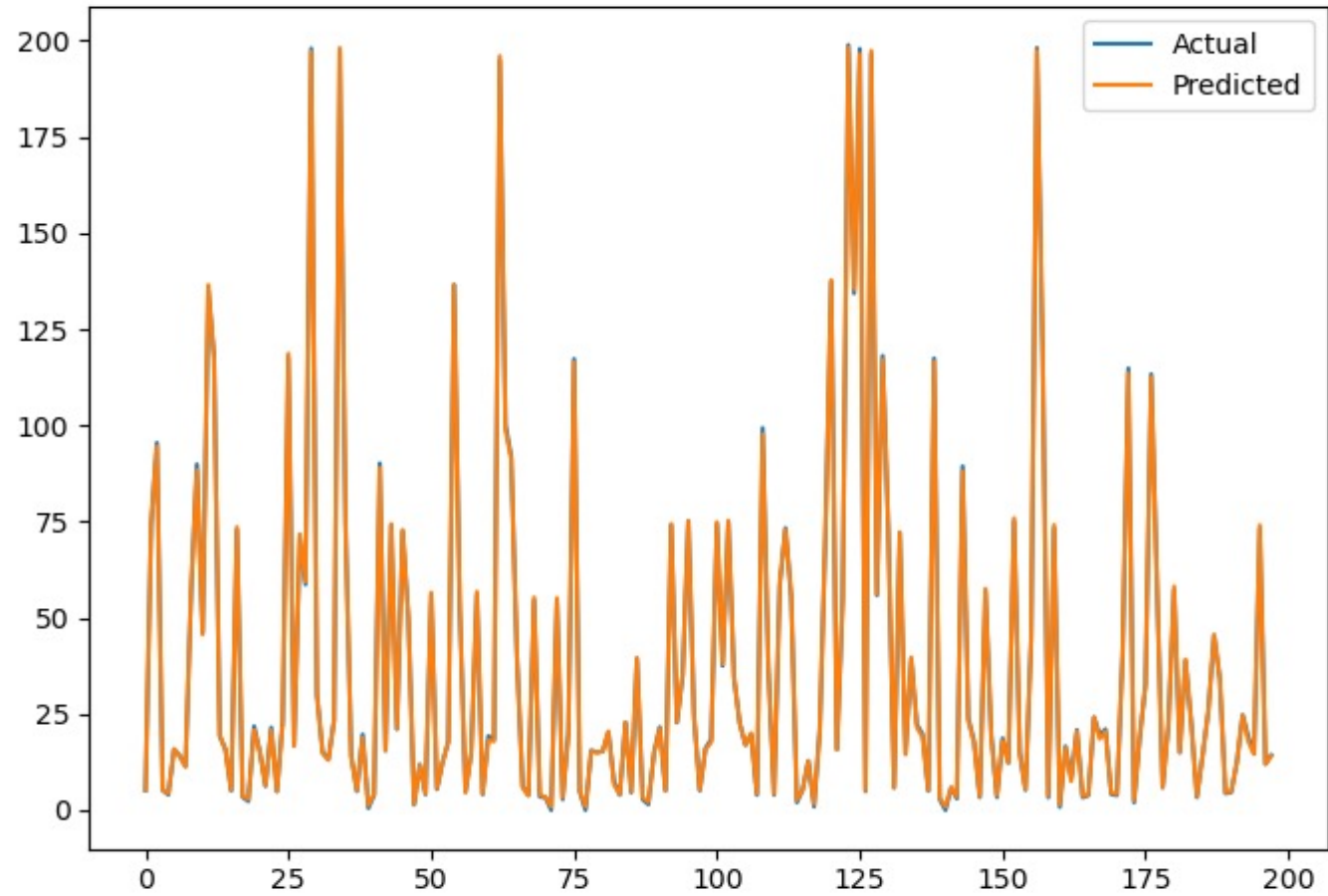
The North Eastern, according to reports is adversely affected by the level of accuracy or lack thereof.

It comprises of

1. Marsabit,
2. Moandera
3. Wajir
4. Garissa



ACTUAL VS PREDICTED RAINFALL



Growth



Shows



Sales

6288816

1.54429012 16.23958709 7.51822795 20.46770983 3.6

7629311

4.05860247 24.12168461 18.52059798 20.46851853 4.5

1261846

4.20427461 41.02084256 113.61758072 2.60985758 19.2

4907672

32.63817096 112.58384853 59.83892559 5.95761141 21.6

6642214

58.16556325 15.05825482 39.22810389 22.60661144 3.6

7531089

14.2574695 25.69443018 45.56312207 34.52591655 4.6

490994

4.83566672 12.42560317 24.60499879 17.87460922 14.5

5071871

74.00000000 11.07112963 14.08530697]

Mean Squared Error: 0.34669606750483895

RESULTS OF THE RBFN MODEL

ACCURACY OF THE MODEL

The RBFN model accurately predicted the long-range weather for North Eastern with an accuracy of 99.7%- $\langle \text{MSE-034666960} \rangle$.

The annual predictions have been plotted against the actual weather for 2013-2023.

The model has predicted the weather for North Eastern for 2024-2034 with an accuracy of 99.7%

Application

These predictions can be used to understand the climate of this region and help in decision-making, planning, and disaster management purposes just to mention but a few.

These predictions can further help the government to improve the economy of this region which will further positively impact Kenya's growing GDP.

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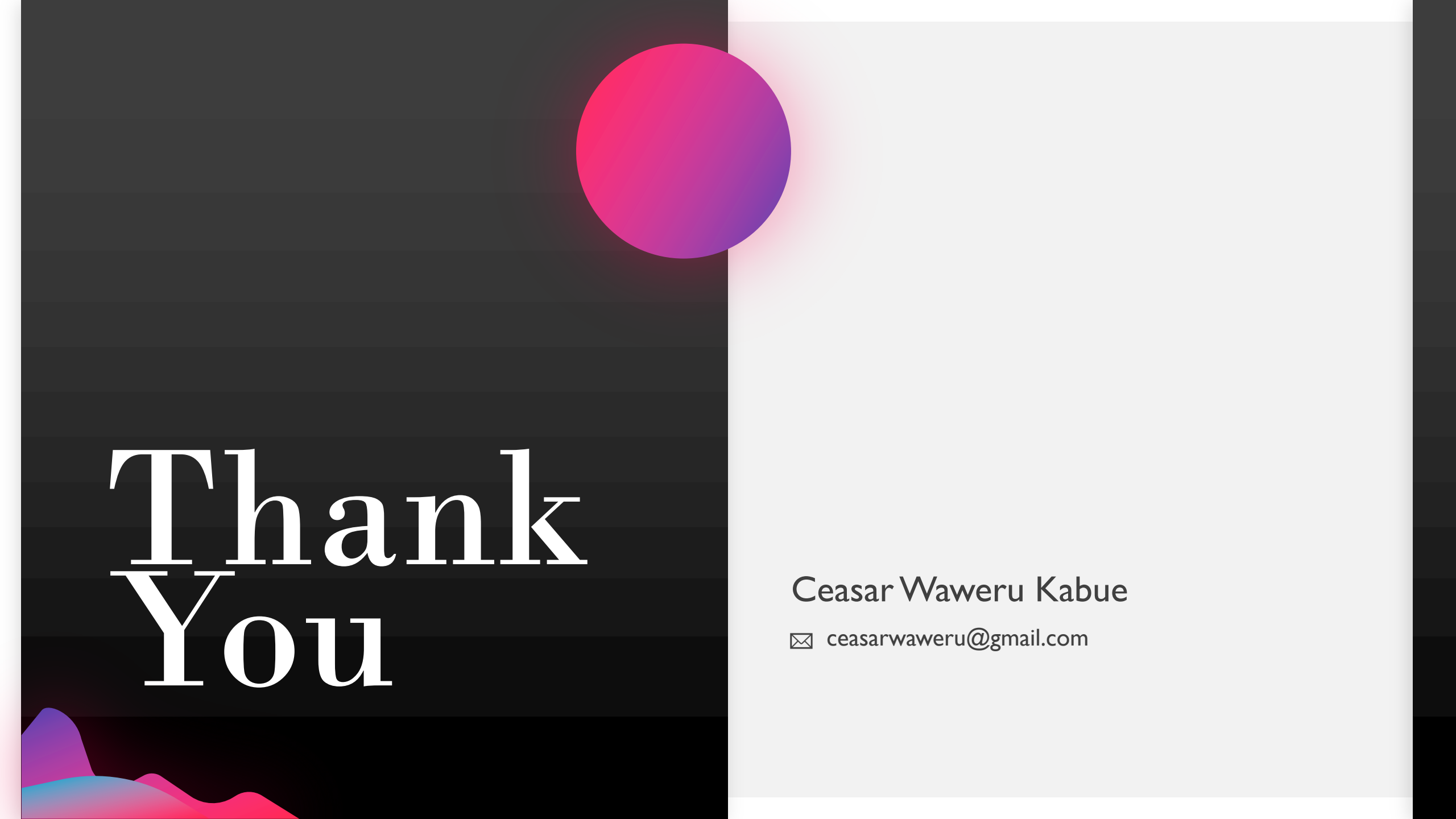
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Thank You

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