

A series of thin, black, overlapping lines forming various geometric shapes like triangles and polygons, scattered across the upper left portion of the slide.

# **FACIAL EMOTION DETECTION**

**CAPSTONE PRESENTATION**

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**12/21/2024**

# AGENDA

PROBLEM DEFINITION

PROBLEM TO DETERMINE

PROPOSED MODEL SOLUTION

RESULTS

FINAL MODEL SOLUTION

PROPOSED BUSINESS  
SOLUTIONs

RECOMMENDATIONS

EXECUTIVE SUMMARY



# PROBLEM DEFINITION

## Problem Definition

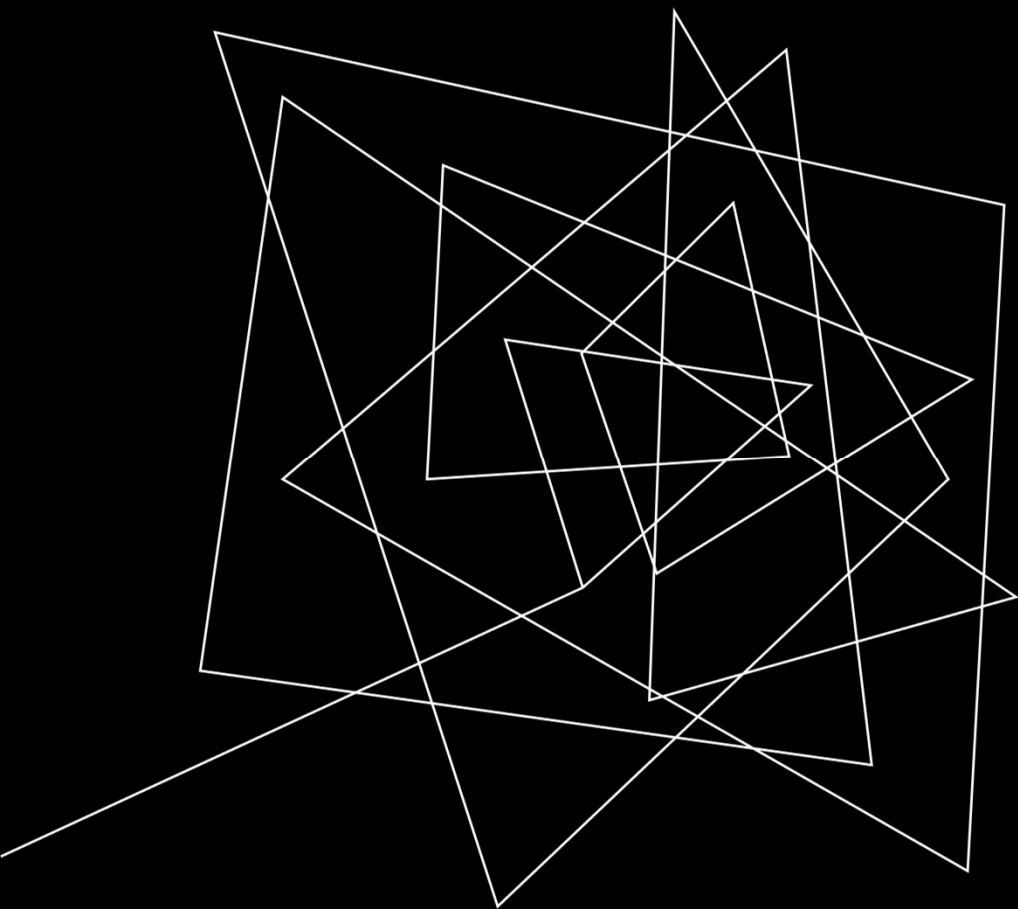
*The objective of this project is to find the most accurate model in order to detect 1 of 4 emotions given digital images of people.*



PROBLEM TO  
DETERMINE

## PROBLEM TO DETERMINE

*To determine which machine learning (ML) model is the most accurate in processing emotional images and if this concept is feasible for a larger implementation in either the business, medical field or beyond.*



PROPOSED MODEL



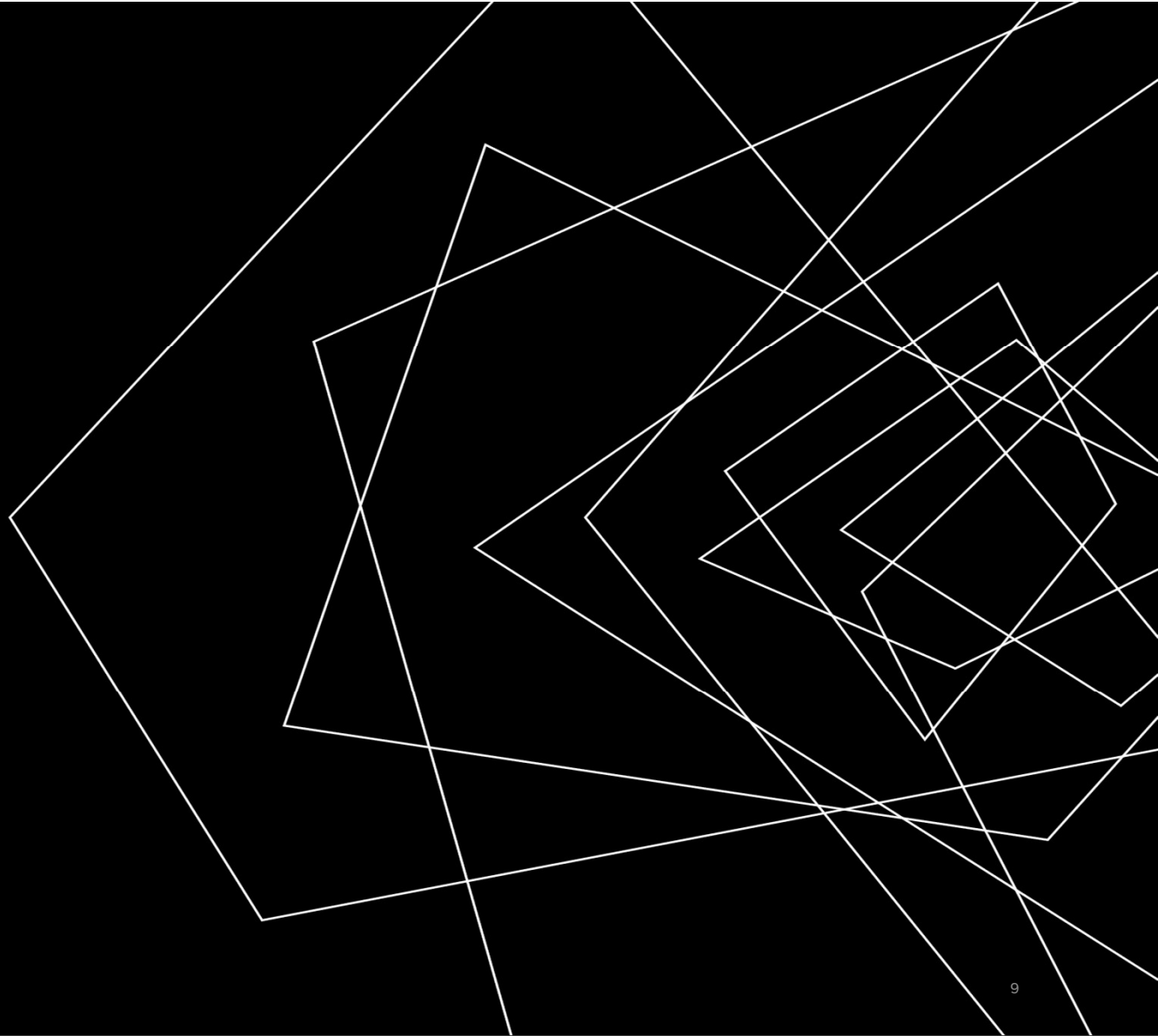
## PROPOSED MODEL

### **Convolutional Neural Network (CNN)**

- Low pre-processing of data
  - Ideal for images
- Translation Invariant
  - Good for identifying facial features
- Scalable



# RESULTS



## RESULTS

Model	Training Accuracy
CNN – model 1	0.7266
CNN – model 2	0.7656
VGG16	0.5312
ResNet V2	0.5546875
EfficientNet	0.5312
5 Convolutional Blocks	0.7188

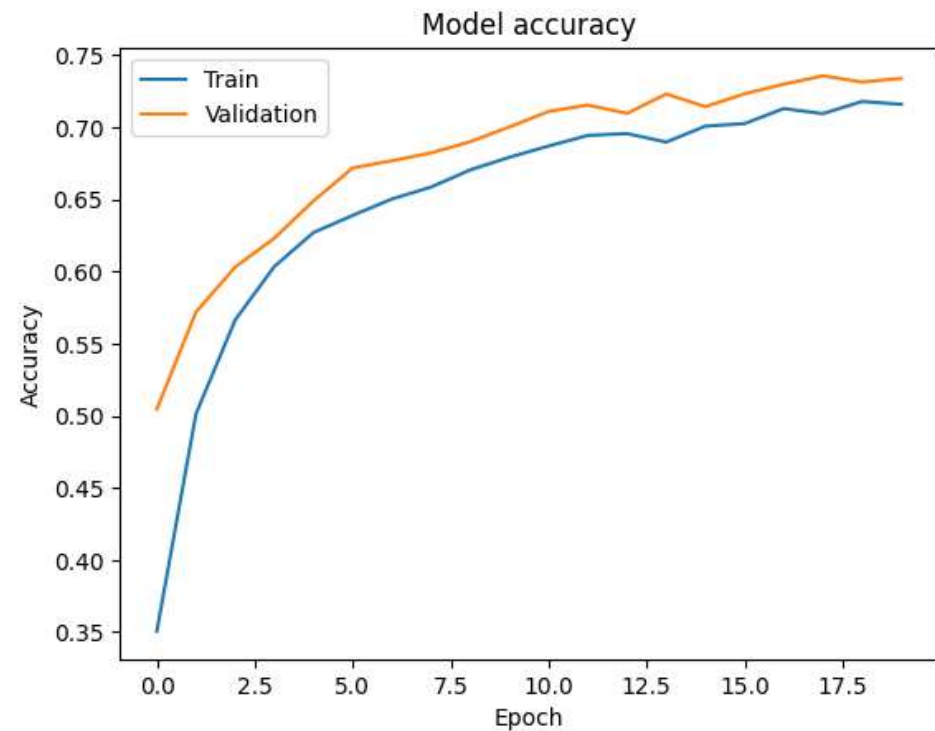


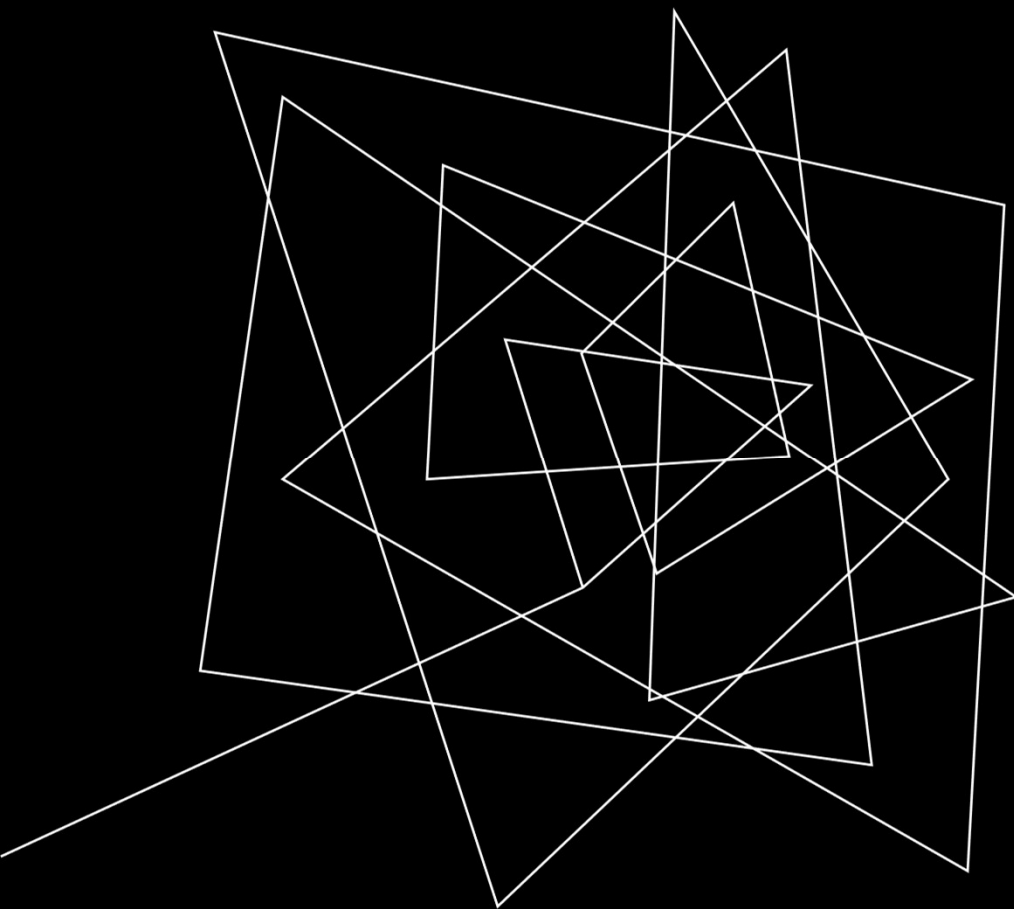
# FINAL MODEL

## FINAL MODEL

### CNN

1. RGB Color mode
2. Relu for activation
3. Pooling (2,2)
4. 0.5 Dropout
5. Softmax activation
6. 20 epochs





# PROPOSED BUSINESS SOLUTIONS



## PROPOSED BUSINESS SOLUTIONS

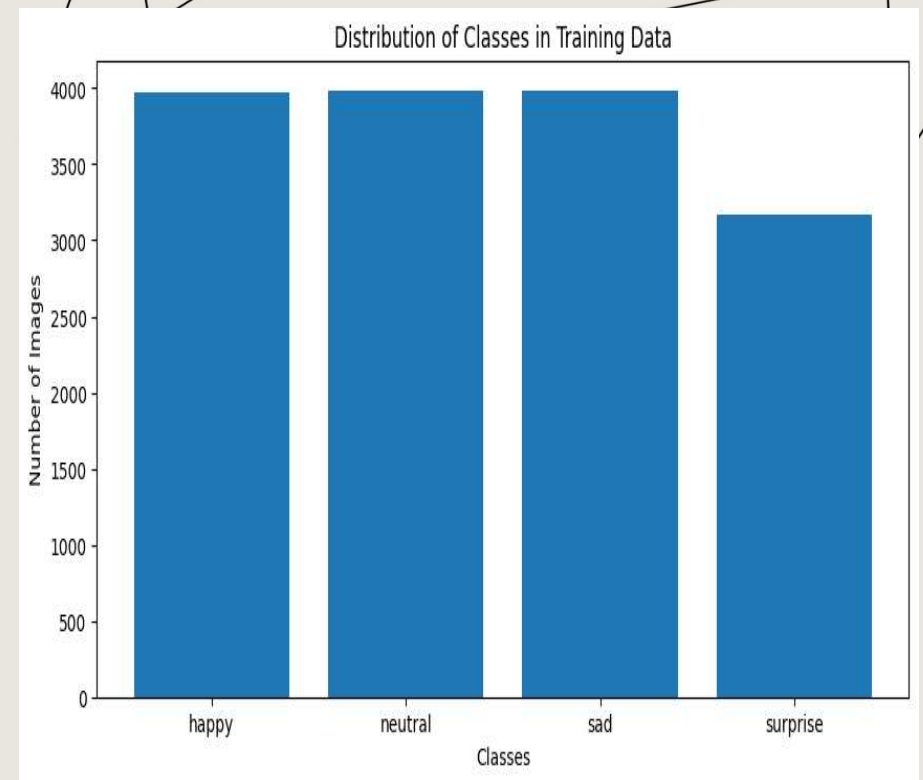
- Retail
  - Account Access
  - AI Assistant evaluation
- Marketing
  - Reaction to advertising/products
- Medical
  - Mental Health diagnosis/treatment
- Information Security
  - Biometrics



# RECOMMENDATIONS

# RECOMMENDATIONS

- Larger data set
- Balanced data set
- Less emphasis on exceptions, more emphasis on typical representations



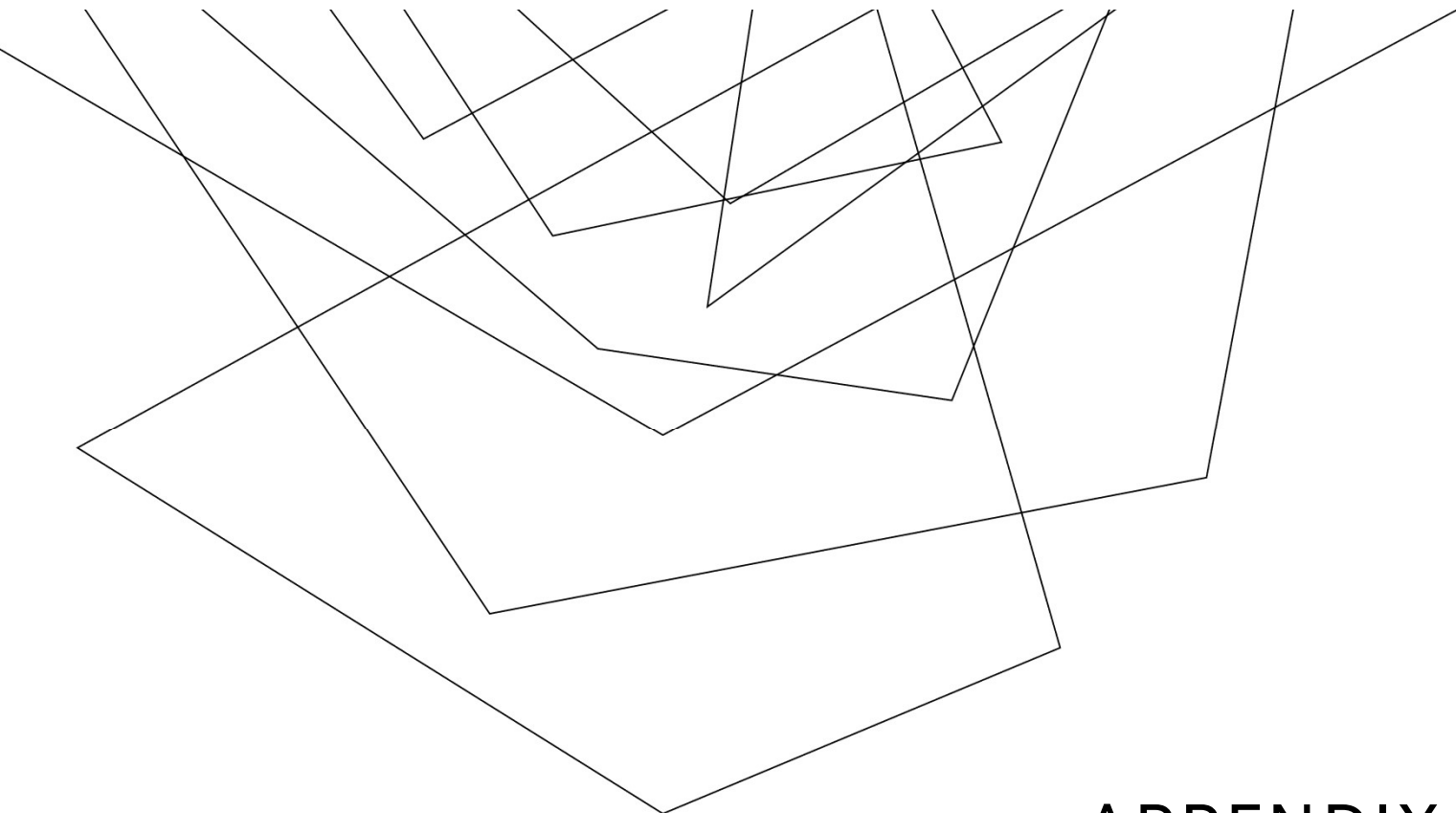
Imbalance existing in data set





# EXECUTIVE SUMMARY

- CNN is the preferred model to use
- Work on balancing out the dataset
- Increase examples from which to train
- Uses in the retail, marketing, medical, and information security industries



APPENDIX

# CODE SAMPLE

+ Code + Text

## ▼ Creating the second Convolutional Neural Network

- Try out a slightly larger architecture

```
[ ] # Define the CNN model2
model2 = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(64, (3, 3), activation='relu', input_shape=(IMG_WIDTH, IMG_HEIGHT, 3 if color_mode == 'rgb' else 1)),
    tf.keras.layers.MaxPooling2D((2, 2)),
    tf.keras.layers.Conv2D(128, (3, 3), activation='relu'),
    tf.keras.layers.MaxPooling2D((2, 2)),
    tf.keras.layers.Conv2D(256, (3, 3), activation='relu'),
    tf.keras.layers.MaxPooling2D((2, 2)),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(256, activation='relu'),
    tf.keras.layers.Dropout(0.5), # Add dropout for regularization
    tf.keras.layers.Dense(4, activation='softmax') # Output layer with 4 classes and softmax activation
])

# Compile the model
model2.compile(optimizer='adam',
               loss='categorical_crossentropy',
               metrics=['accuracy'])

# Print the model summary
model2.summary()
```

## CNN MODEL 2