


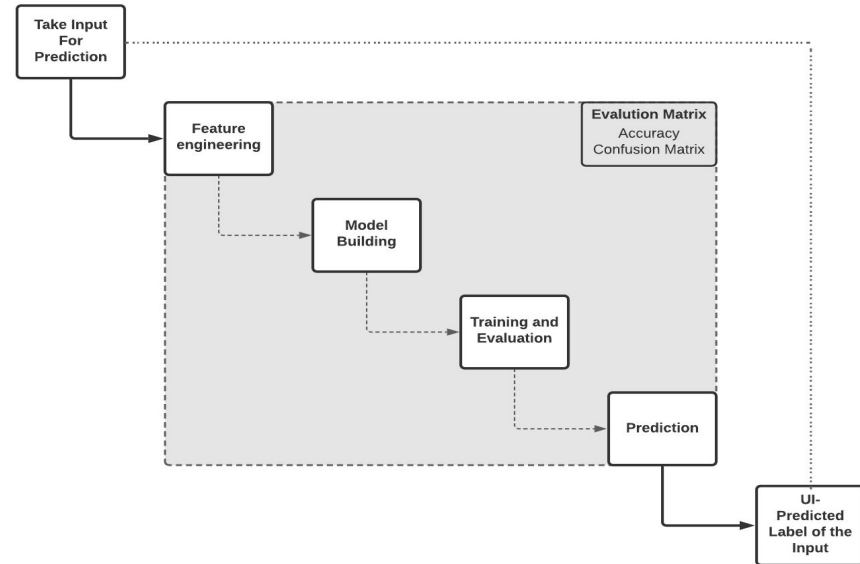
Letter Prediction

Cory Randolph, Abraham Kong,
Akanksha Rawat, Karishma Kuria

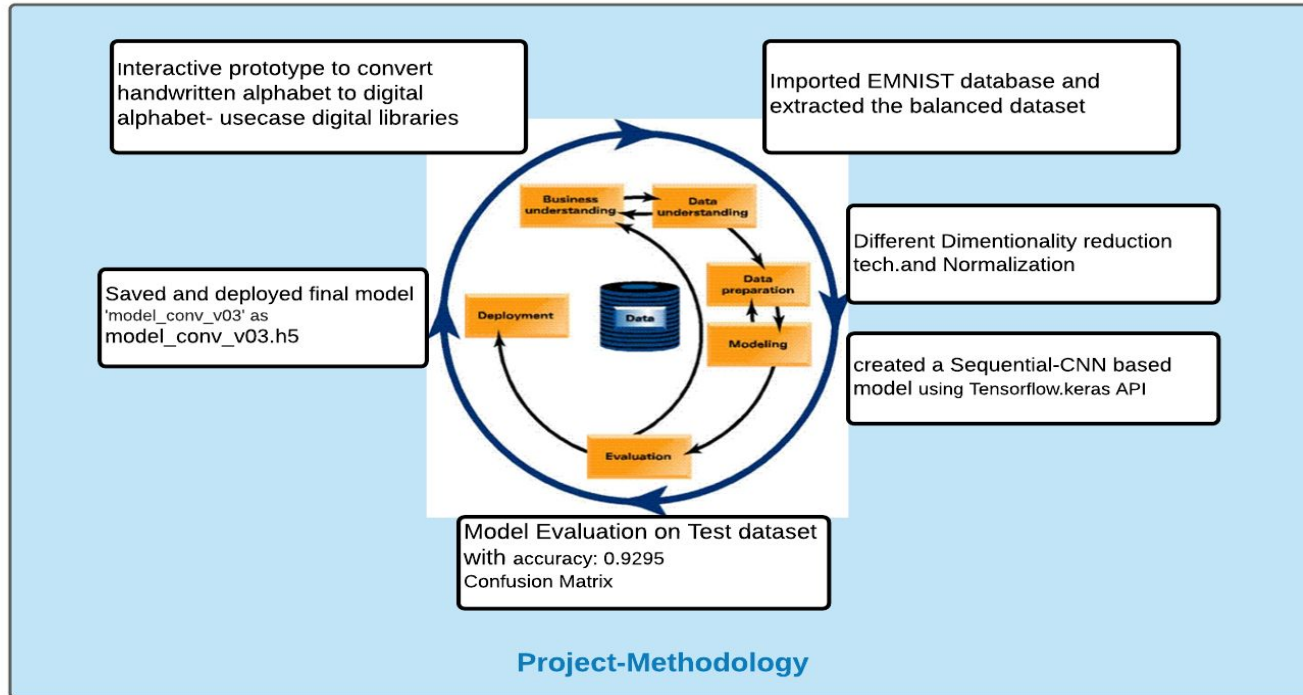
A dark blue diagonal gradient bar that starts from the bottom left and extends towards the top right, covering the lower half of the slide.

Abstract

- Build an whiteboard on a front end UI
- Take user drawn had written letters
- Apply Machine Learning Model to classify drawing
- Return a prediction based on the user's drawn letter



Methodology



Data Understanding

- Imported EMNIST- balanced dataset.
- The EMNIST dataset has a set of handwritten digits (0–9), (a-z), and (A-Z) being converted into 28x28 pixel pictures.
- The EMNIST Balanced dataset - 47-class dataset was chosen over the By Class dataset to avoid classification errors.

Data Preprocessing

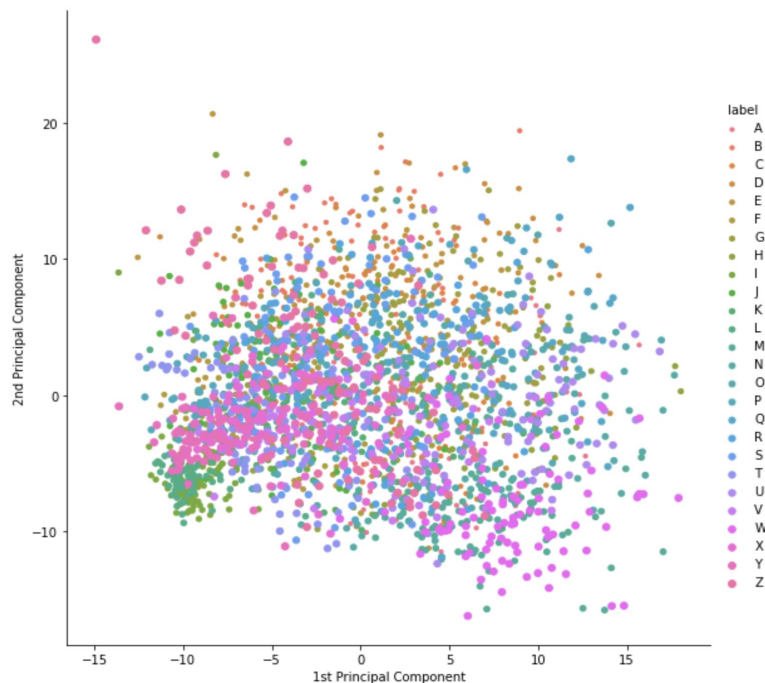
1. **Data Reduction Techniques used:**

- Principal Component Analysis (PCA)
- Singular Value Decomposition (SVD)
- Locally Linear Embedding (LLE)
- T-distributed Stochastic Neighbor Embedding (T-SNE)
- Isometric Mapping (ISOMAP)
- Uniform Manifold Approximation and Projection (UMAP)

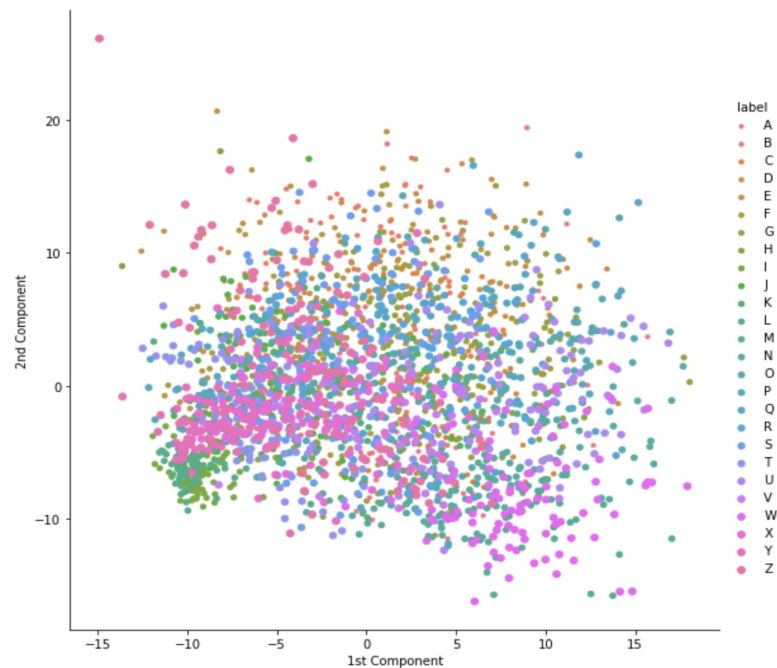
2. **Data Normalization:** The data set is divided by 255 based on the RGB codes.

Data Reduction Visualizations

PCA

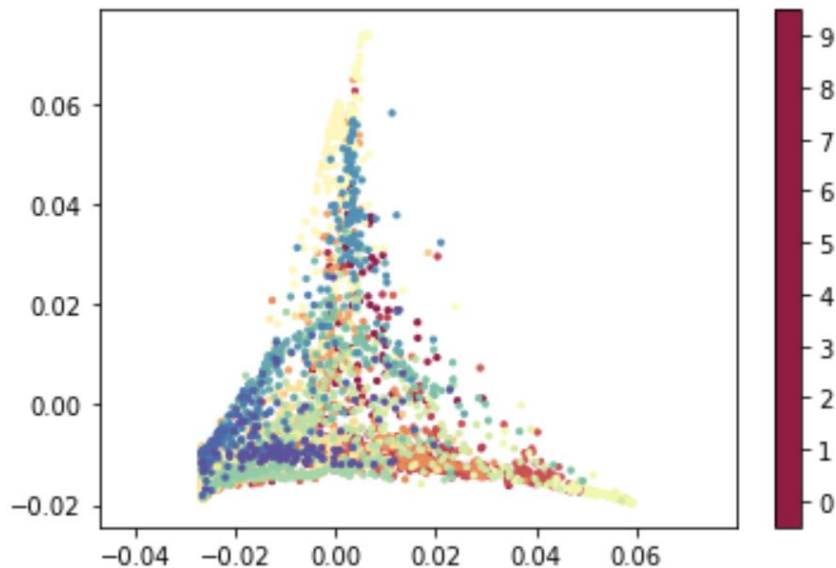


SVD

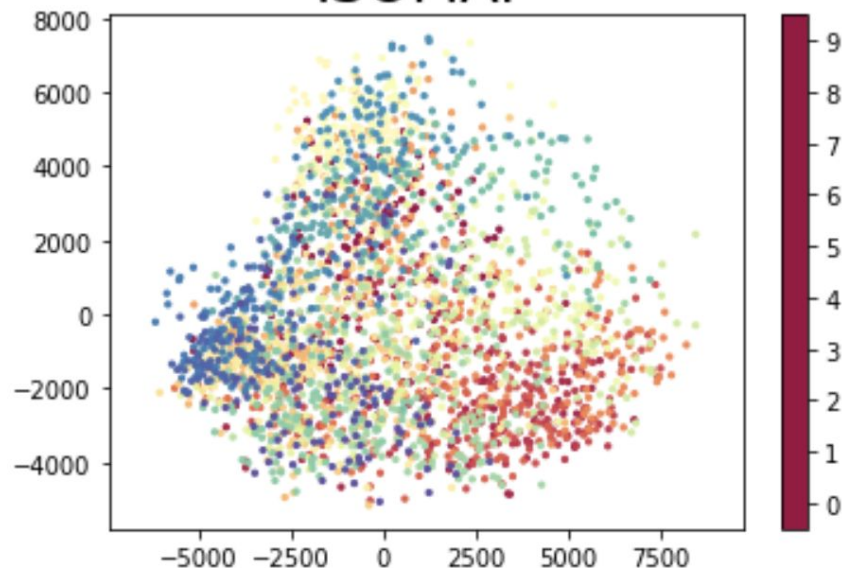


Data Reduction Visualizations

LLE

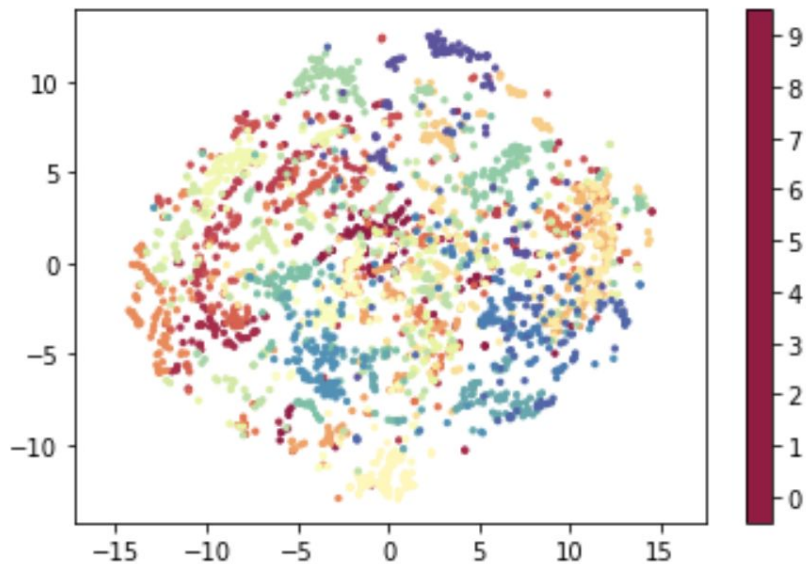


ISOMAP

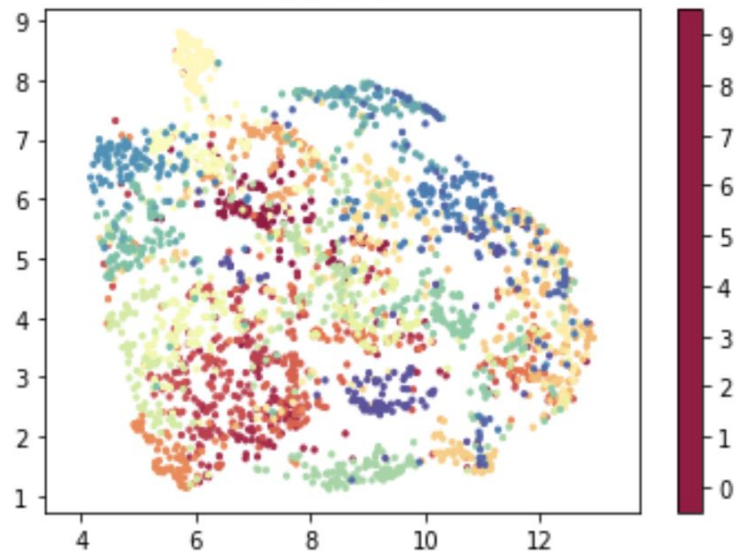


Data Reduction Visualizations

t-SNE



UMAP



Modeling

1. **Building the model-** Using Keras- high level API of tensorflow framework build a layered sequential type

Model.- Based on **Convolutional Neural Network architecture**. model type= Sequential()

```
model.add(Conv2D(32, kernel_size=3, activation='relu', input_shape=(28, 28, 1)))
```

2. **Compiling the model-** parameters [optimizer, Loss, Metrics]

```
model.compile(optimizer='adam',  
              loss='categorical_crossentropy',  
              metrics=['accuracy'])
```

```
model = Sequential()
```

```
model.add(Conv2D(32, kernel_size=3, activation='relu', input_shape=(28, 28, 1)))  
model.add(BatchNormalization())  
model.add(Conv2D(32, kernel_size=3, activation='relu'))  
model.add(BatchNormalization())  
model.add(Conv2D(32, kernel_size=5, strides=2, padding='same', activation='relu'))  
model.add(BatchNormalization())  
model.add(Dropout(0.4))
```

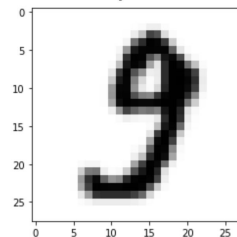
3. **Training the Model**

```
history = model.fit(x = x_train,  
                    y = y_train,  
                    validation_data = (x_test, y_test),  
                    epochs=20) #10
```

4. **Evaluation and Prediction**

```
model.evaluate(x_test, y_test)
```

Predicted Label: g
Actual Label: g



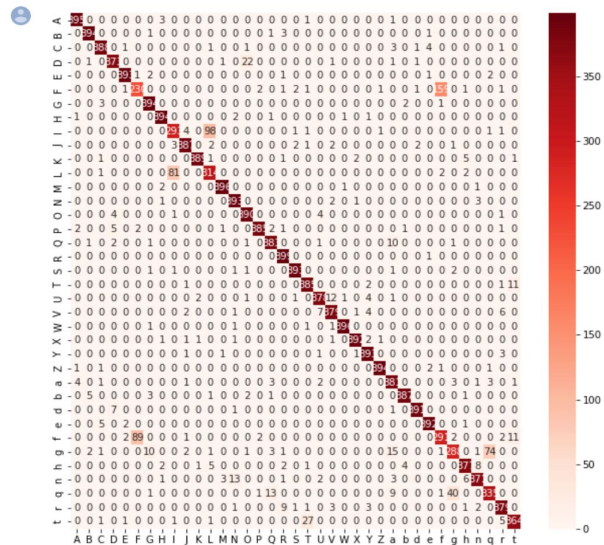
Evaluation

Evaluated on test data set. Using below measures:

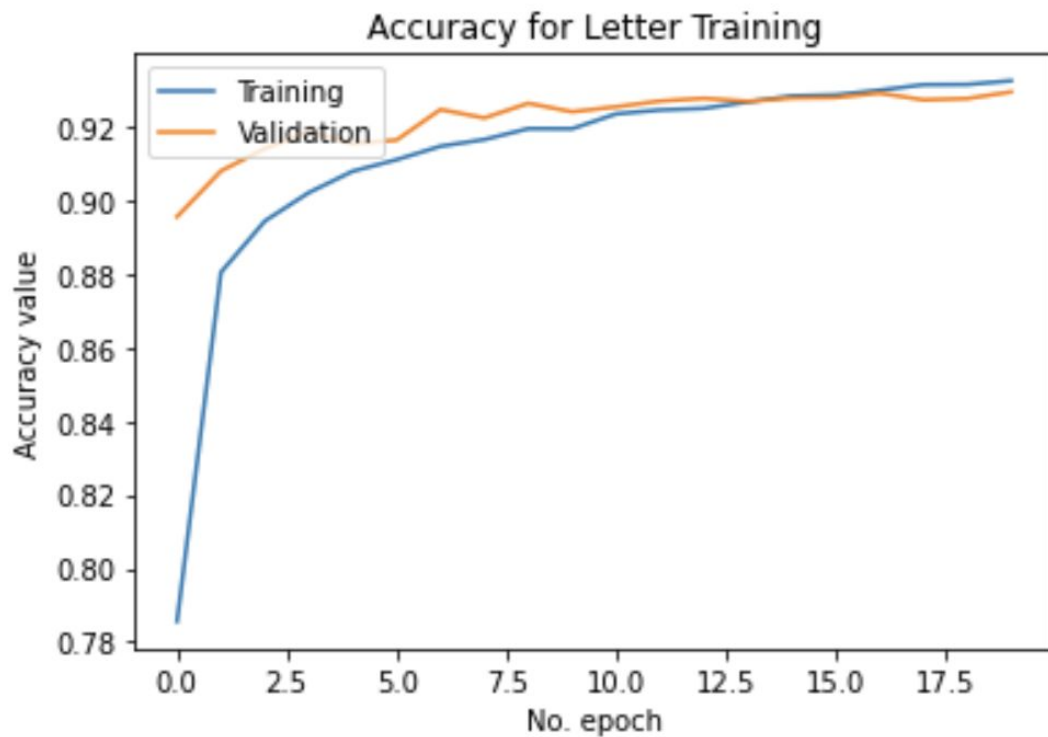
- Confusion Matrix
- Accuracy
- Loss
- Precision
- Recall
- f1-score

```
[ ] cm = confusion_matrix(y_test_as_labels, y_test_preds_as_labels)
```

```
plt.figure(figsize=[10,10])  
sns.heatmap(cm, cmap="Reds", annot=True, fmt='.0f', xticklabels = labels_dict_caps.values(), yticklabels= labels_dict_caps.values())  
plt.show()
```

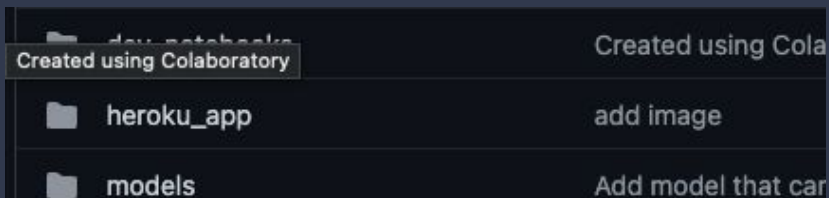


Evaluation



Deployment: Heroku

We choose Heroku as our web server
To deploy our app, as it is very easy to
Deploy once the Github is set right.



Deployment method



App connected to GitHub

Code diffs, manual and auto deploys are
available for this app.

Connected to [AbrahamKong/ernist_letter_exploration_and_prediction](#) by [AbrahamKong](#)

Disconnect...

Releases in the [activity feed](#) link to GitHub to view commit diffs

Deployment: Heroku

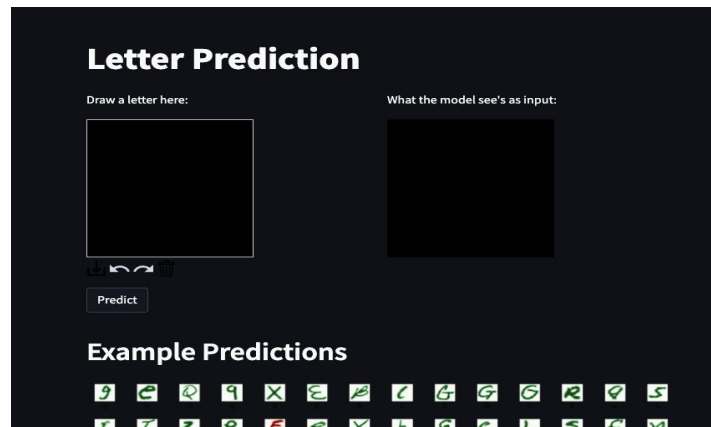
reference	add image
.gitignore	update requirements
Procfile	update
README.md	Update Heroku setup readme
app.py	Add example training predictions
requirements.txt	Update requirements
setup.sh	add heroku files

- app.py generated by our Google Colab file
Including our Streamlit Front end and modeling
- Requirements.txt and setup.sh to set up the environment
- Procfile to tell Heroku to run the file

Deployment Link:

<https://letter-prediction.herokuapp.com/>

Will introduce more in the live demo part coming up



Front End Design (Streamlit)

Main Features

- User draws a letter
- Custom ML model makes a class/letter prediction
- Probability of labels displayed

```
with col1:
    st.markdown('Draw a letter here:')
    # Create a drawing canvas with desired properties
    canvas_result = st_canvas(
        fill_color="#ffffff",
        stroke_width=10,
        stroke_color='#ffffff',
        background_color="#000000",
        height=200,
        width=200,
        drawing_mode='freedraw',
        key="canvas",
    )

with col2:
    # Show that the resized image looks like
    st.markdown("What the model see's as input:")
    if canvas_result.image_data is not None:
        img = cv2.resize(canvas_result.image_data.astype('uint8'), (28, 28))
        img_rescaling = cv2.resize(img, (200, 200), interpolation=cv2.INTER_NEAREST)
        st.image(img_rescaling)

# Generate the prediction based on the users drawings
if st.button('Predict'):
    x_user_input = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    x_user_input = x_user_input.reshape(1, 28, 28, 1) / 255 # Reshape and normalize
    pred = model.predict(x_user_input)
    pred_label = labels_dict[pred.argmax()]
    st.header(f'Predicted Label: {pred_label}')

# Create a Plotly barchart of the predicted probabilities
fig = create_probability_fig(pred)
st.plotly_chart(fig, use_container_width = True)
```

Live Demo

Heroku Hosted Demo:

<https://letter-prediction.herokuapp.com/>

Colab Demo:

https://github.com/coryroyce/emnist_letter_exploration_and_prediction/blob/main/streamlit_application/Streamlit_App.ipynb

* Can be run even if the demo app is no longer being actively hosted

Letter Prediction

Draw a letter here:

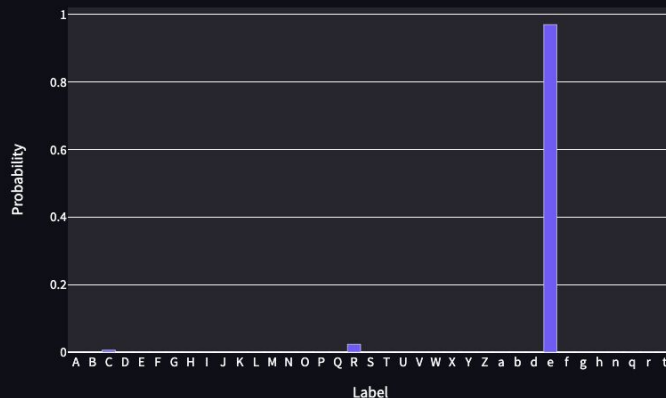


Predict

What the model see's as input:



Predicted Label: e



Thank you!

- Cory Randolph, Abraham Kong,
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