# Diabetic Foot Exam System

Old Dominion University Fall 2021 Stephanie Trusty



#### **Table of Contents**

- 1. Problem Description
  - a. Background
  - b. Project Goals
  - c. Dataset
  - d. Limitations and Constraints
- 2. Solution and Implementation
  - a. Algorithm Flow
  - b. Accuracy and Loss
  - c. Test Cases
- 3. Future Considerations
  - a. Recommendations for Improvement
- 4. Conclusion

# **Problem Description**

### What is a Diabetic Foot Exam?

A 3-minute professional evaluation to identify a diabetic patient's risk of foot ulceration. Exam components include:

- Patient history
- Dermatological assessment
- Musculoskeletal assessment
- Neurological assessment

#### **Project Goals**

#### **Dermatological Assessment**

- Calluses
- Blisters

#### Musculoskeletal Assessment

- Foot Deformities
  - Clawtoe
  - Hammertoe
  - Bunion



### Project Equipment

- Raspberry Pi 3 Model B
- Raspberry Pi Camera Module v2
- Keyboard
- Mouse



### **Python Libraries and Packages**

- PiCamera
  - Controlling the Raspberry Pi Camera Module
- Matplotlib
  - Python plotting library
- Python Imaging Library
  - Open, save, rotate image files
- TensorFlow
  - Software library for machine learning and AI

#### Image Dataset

- The dataset consists of 420 images placed into three categories.
  - Callus
  - Blister
  - Deformation
- Training images: 336
- Validation images: 84

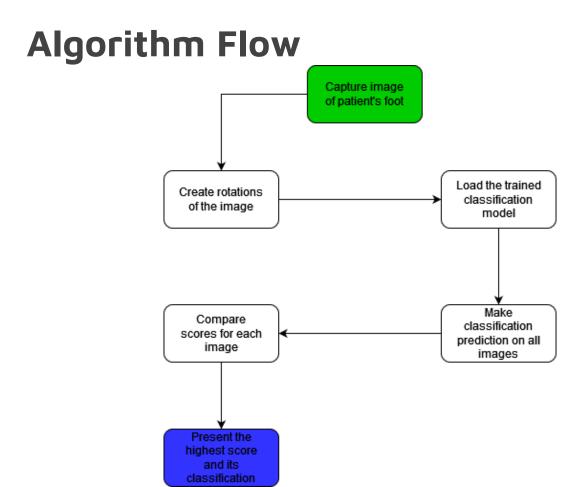
### **Limitations and Constraints**

- Limited Dataset Size
  - Ideal size: 1,000 images per class
  - Actual size: 140 images per class
- Lack of Diversity in Data
  - $\circ \qquad {\sf Skin \ tone} \qquad$
  - Foot Placement
  - Aging Skin
- Address the most predominant condition

Solution and Implementation

### **Implementation Challenges**

- TensorFlow version and Raspberry Pi compatibility
- Data uniqueness
- Dataset selection
- Camera position and lighting



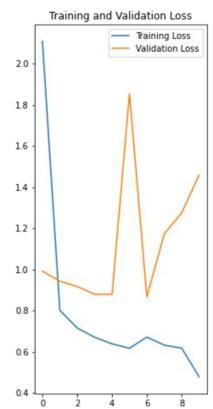
### **Training and Validation Loss**

#### **Training Comparison**

- Training Loss (start): 2.1089
- Training Loss (end): 0.4783

#### Validation Comparison

- Validation Loss (start): 0.9912
- Validation Loss (end): 1.5483



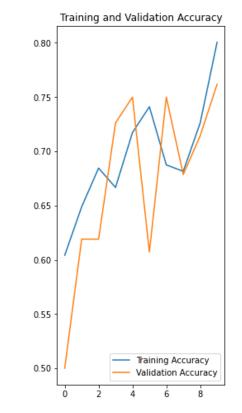
### **Training and Validation Accuracy**

#### **Training Comparison:**

- Training Accuracy (start): 0.6042
- Training Accuracy (end): 0.8006

#### Validation Comparison

- Validation Accuracy (start): 0.5000
- Validation Accuracy (end): 0.7619



#### Loss and Overfitting

Total params: 47,341,091 Trainable params: 47,341,091 Non-trainable params: 0

Epoch 1/10 Epoch 2/10 Epoch 3/10 34/34 [=========== ======] - 146s 4s/step - loss: 0.7147 - accuracy: 0.6845 - val loss: 0.9164 - val accuracy: 0.6190 Epoch 4/10 34/34 [=========== ======] - 146s 4s/step - loss: 0.6700 - accuracy: 0.6667 - val loss: 0.8786 - val accuracy: 0.7262 Epoch 5/10 ============] - 147s 4s/step - loss: 0.6380 - accuracy: 0.7173 - val loss: 0.8786 - val accuracy: 0.7500 Epoch 6/10 34/34 [====== ===] - 146s 4s/step - loss: 0.6167 - accuracy: 0.7411 - val loss: 1.8535 - val accuracy: 0.6071 Epoch 7/10 34/34 [======= ===] - 147s 4s/step - loss: 0.6706 - accuracy: 0.6875 - val loss: 0.8663 - val accuracy: 0.7500 Epoch 8/10 Epoch 9/10 34/34 [===============================] - 147s 4s/step - loss: 0.6168 - accuracy: 0.7262 - val loss: 1.2744 - val accuracy: 0.7143 Epoch 10/10 

#### Evidence of some overfitting

### Case Study: Raspberry Pi Image

#### **Bottom View**

- Prediction: Deformation
- Confidence: 84.54%
- Correct classification: Callus

#### Side View

- Prediction: Deformation
- Confidence: 96.45%
- Correct classification: Callus





#### Case Study Results: Dermatological





**Prediction:** Deformation

**Confidence:** 92.36%

Correct Classification: Callus

Prediction: Callus

Confidence: 80.73%

Correct Classification: Callus

#### Case Study Results: Musculoskeletal



**Prediction:** Deformation

Confidence: 78.41%

Correct Classification: Deformation

Prediction: Callus

**Confidence:** 99.62%

Correct Classification: Deformation



# **Future Considerations**

### **Recommendations for Improvement**

- Build data set with original images
- Verify data for uniqueness
  - Ensure each image is unique
- Separate deformation categories
  - Determine more accurate characteristics for a class
- Address patients with multiple areas of concern

### Conclusion

- Inaccurate results with high confidence levels
- Evidence of overfitting
- Larger, more diverse dataset is needed
- Additional study needed to determine impact of lighting and camera position

### References

- <u>https://care.diabetesjournals.org/content/31/8/1679</u>
- <u>https://diabetesed.net/wp-content/uploads/2017/05/3-minute-foot-exam.pdf</u>
- <u>https://www.ibm.com/cloud/learn/overfitting</u>
- <u>https://www.tensorflow.org/</u>
- <u>https://keras.io/</u>