

# **PENGEMBANGAN MODEL KLASIFIKASI MONKEYPOX MENGGUNAKAN CNN TRANSFER LEARNING VGG16 DENGAN OPTIMIZER ADAM DAN AKSELERASI CUDA**

**Imam Afandy 21081010290**

# RUMUSAN MASALAH

- Bagaimana meningkatkan akurasi dan efisiensi deteksi monkeypox menggunakan transfer learning?
- Bagaimana penggunaan arsitektur VGG16 dan akselerasi CUDA dapat membantu mengatasi keterbatasan data dan waktu komputasi?

# RESEARCH GAP

- Sebagian besar penelitian sebelumnya fokus pada penyakit kulit umum dan penyakit infeksi lain, tetapi masih sedikit penelitian tentang klasifikasi otomatis monkeypox menggunakan deep learning.
- kurangnya dataset besar dan model khusus yang dipotimalkan untuk klasifikasi monkeypox.

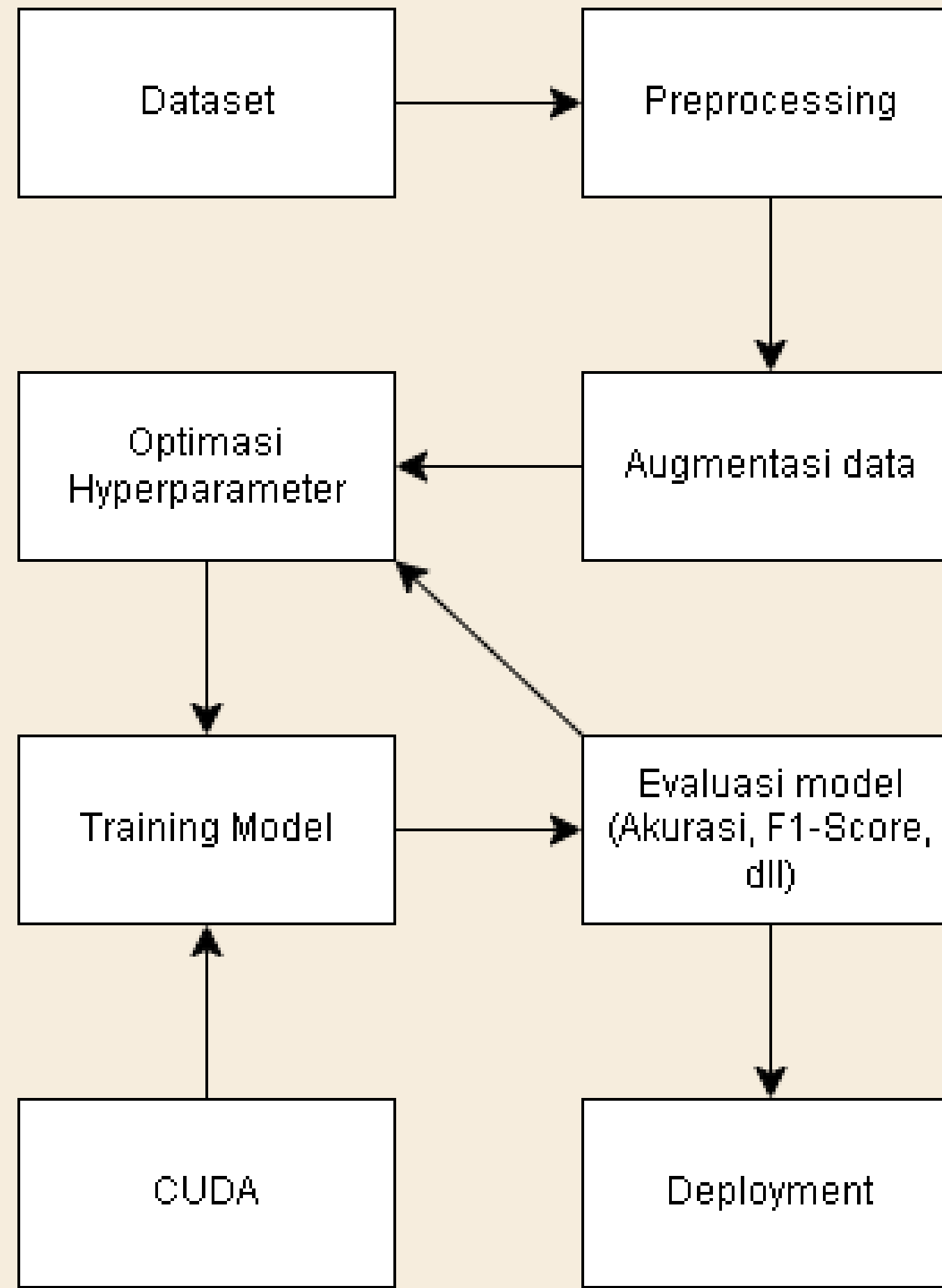
# MIND MAPPING REFERENSI

Monkeypox Detection Using CNN with Transfer Learning (Murat Altun 1 , Hüseyin Gürüler 1 , Osman Özkaraca 1 , Faheem Khan 2,\* , Jawad Khan 3 and Youngmoon Lee 3,\*

1)

- 1.Dataset : Gambar lesi kulit dari dataset penelitian terdahulu
  - 2.Model : Pre-Trained VGG16 sebagai feature extractor
  - 3.Optimizer : Adam untuk efisiensi dan akurasi
  - 4.Implementasi CUDA :  
Mempercepat proses training dan validasi
- Output : Model Klasifikasi dengan metrik seperti akurasi, F1-Score, dan waktu komputasi.

# Metodologi





```
1 import tkinter as tk
2 from tkinter import filedialog
3 from PIL import Image, ImageTk
4 from tensorflow.keras.models import load_model
5 import numpy as np
6
7 # Muat model
8 model = load_model('Model/my_model.h5')
9
10 # Fungsi untuk memprediksi gambar
11 def predict_image(img_path):
12     img = Image.open(img_path)
13     img = img.resize((224, 224))
14     img_array = np.array(img, dtype=np.float32) # Ubah tipe data ke float32
15     img_array = np.expand_dims(img_array, axis=0) # Menambahkan dimensi batch
16     img_array /= 255.0 # Normalisasi
17
18     # Memprediksi
19     prediction = model.predict(img_array)
20     if prediction[0] > 0.5:
21         return "Monkeypox detected"
22     else:
23         return "Normal Skin"
24
25 # Fungsi untuk menangani klik tombol
26 def upload_image():
27     file_path = filedialog.askopenfilename(filetypes=[("Image Files", "*.jpg;*.jpeg;*.png")])
28     if file_path:
29         result = predict_image(file_path)
30         result_label.config(text=result)
31
32     # Tampilkan gambar
33     try:
34         img = Image.open(file_path)
35         img = img.convert('RGB') # Pastikan gambar dalam format RGB
36         img = img.resize((150, 150))
37         img_tk = ImageTk.PhotoImage(img)
38         img_label.config(image=img_tk)
39         img_label.image = img_tk
40     except Exception as e:
41         result_label.config(text=f"Error loading image: {e}")
42
43 # Membuat UI
44 root = tk.Tk()
45 root.title("Monkeypox Detector")
46
47 upload_button = tk.Button(root, text="Upload Image", command=upload_image)
48 upload_button.pack(pady=20)
49
50 img_label = tk.Label(root)
51 img_label.pack(pady=10)
52
53 result_label = tk.Label(root, text="", font=("Helvetica", 16))
54 result_label.pack(pady=10)
55
56 root.mainloop()
57
```



```
1 import tensorflow as tf
2 from tensorflow.keras.applications import VGG16
3 from tensorflow.keras.preprocessing.image import ImageDataGenerator
4 from tensorflow.keras.layers import Dense, Flatten, Dropout
5 from tensorflow.keras.models import Sequential
6 from tensorflow.keras.optimizers import Adam
7 import matplotlib.pyplot as plt
8 from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
9 import numpy as np
10
11 # Augmentasi data lebih lanjut
12 datagen = ImageDataGenerator(
13     rescale=1./255,
14     rotation_range=40,
15     width_shift_range=0.2,
16     height_shift_range=0.2,
17     shear_range=0.2,
18     zoom_range=0.2,
19     horizontal_flip=True,
20     fill_mode='nearest',
21     validation_split=0.2
22 )
23
24 # Memuat data training dan validasi
25 train_data = datagen.flow_from_directory(
26     'G:/My Drive/SKRIPSI/2/Dataset/Monkeypox Skin Image Dataset',
27     target_size=(224, 224),
28     batch_size=32,
29     class_mode='binary',
30     subset='training'
31 )
32
33 validation_data = datagen.flow_from_directory(
34     'G:/My Drive/SKRIPSI/2/Dataset/Monkeypox Skin Image Dataset',
35     target_size=(224, 224),
36     batch_size=32,
37     class_mode='binary',
38     subset='validation'
39 )
40
41 # Load VGG16 tanpa fully connected layer teratas
42 vgg16_model = VGG16(weights='imagenet', include_top=False, input_shape=(224, 224, 3))
43
44 # Membekukan layer VGG16
45 for layer in vgg16_model.layers:
46     layer.trainable = False
47
48 # Membuat model Sequential dengan dropout
49 model = Sequential()
50 model.add(vgg16_model)
51 model.add(Flatten())
52 model.add(Dense(256, activation='relu'))
53 model.add(Dropout(0.5)) # Dropout layer untuk regularisasi
54 model.add(Dense(1, activation='sigmoid'))
55
56 # Compile model dengan learning rate yang lebih kecil
57 model.compile(optimizer=Adam(lr=0.00001),
58               loss='binary_crossentropy',
59               metrics=['accuracy'])
60
61 # Train model
62 history = model.fit(
63     train_data,
64     epochs=50,
65     validation_data=validation_data
66 )
67
68 # Evaluasi model
69 val_loss, val_acc = model.evaluate(validation_data)
70 print(f"Validation accuracy: {val_acc:.4f}")
71
72 # Plotting Akurasi dan Loss
73 acc = history.history['accuracy']
74 val_acc = history.history['val_accuracy']
75 loss = history.history['loss']
76 val_loss = history.history['val_loss']
77
78 epochs_range = range(1, len(acc) + 1)
79
80 plt.figure(figsize=(12, 6))
81
82 # Plot Akurasi
83 plt.subplot(1, 2, 1)
84 plt.plot(epochs_range, acc, label='Training Accuracy')
85 plt.plot(epochs_range, val_acc, label='Validation Accuracy')
86 plt.legend(loc='lower right')
87 plt.title('Training and Validation Accuracy')
88
89 # Plot Loss
90 plt.subplot(1, 2, 2)
91 plt.plot(epochs_range, loss, label='Training Loss')
92 plt.plot(epochs_range, val_loss, label='Validation Loss')
93 plt.legend(loc='upper right')
94 plt.title('Training and Validation Loss')
95
96 plt.show()
97
98 # Menyimpan model
99 model.save('Model/my_model.h5')
100
101 # Menambahkan confusion matrix
102 # Mengambil prediksi dari model
103 validation_labels = validation_data.classes
104 validation_predictions = (model.predict(validation_data) > 0.5).astype("int32")
105
106 # Menghitung confusion matrix
107 cm = confusion_matrix(validation_labels, validation_predictions)
108
109 # Menampilkan confusion matrix
110 disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=validation_data.class_indices.keys())
111 disp.plot(cmap=plt.cm.Blues)
112 plt.title('Confusion Matrix')
113 plt.show()
114
```

# PROGRESS PENGERJAAN



**THANK YOU**