

## **BAB VI.**

## **PENUTUP**

### **6.1.Kesimpulan**

Berdasarkan hasil analisis, pembahasan, dan pengujian, dapat disimpulkan bahwa sistem yang dibangun berhasil mengimplementasikan *Deep Learning* menggunakan *Convolutional Neural Network* (CNN) dengan menggunakan arsitektur MobileNetV2 untuk mengidentifikasi jenis-jenis bunga. Selain itu, sistem juga berhasil dibangun dalam bentuk aplikasi *mobile* berbasis iOS yang dapat mengidentifikasi jenis-jenis bunga.

### **6.2.Saran**

Beberapa saran yang dapat diberikan penulis untuk penelitian dan pembangunan sistem selanjutnya adalah sebagai berikut:

1. Dataset dapat dibuat terhubung dengan internet agar dapat memungkinkan dilakukannya pembaharuan dataset. Supaya jika ada data bunga baru, sistem dapat menerima data tersebut untuk klasifikasi.

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## Lampiran

**Gambar 1. Confusion Matrix Model 1**

**Gambar 2. Confusion Matrix Model 2**

**Gambar 3. Confusion Matrix Model 3**

**Gambar 4. Confusion Matrix Model 4**

Gambar 5. Confusion Matrix Model 5

