

BAB VI

KESIMPULAN DAN SARAN

6.1 Kesimpulan

Berdasarkan hasil dan pembahasan di atas, maka penulis dapat merumuskan beberapa kesimpulan sebagai berikut :

1. Model algoritma *Neural Network* dan *Random Forest* dapat direkomendasikan untuk melakukan prediksi hasil pertandingan *Mobile legends bang bang* (MLBB) dengan masing – masing akurasi sebesar 82.30 % dan 80.04 %.
2. Dalam perancangan model algoritma *Neural Network*, jumlah persentase data latih, jumlah neuron dalam *hidden layer*, dan jumlah *epoch* yang digunakan dapat mempengaruhi hasil skor akurasi model.
3. Jumlah node dalam perancangan model algoritma *Random Forest* dapat mempengaruhi hasil skor akurasi model.

6.2 Saran

Penulisan ini hanya berfokus pada *game Mobile legends bang bang* (MLBB), dimana dataset yang digunakan merupakan hasil histori pertandingan pada tier *mythic*. Untuk saran bagi penulisan selanjutnya, dapat mencoba menambah fitur yang digunakan dalam penulisan, seperti jenis *hero* yang di gunakan pemain, ataupun faktor – faktor lain yang mempengaruhi hasil pertandingan *Mobile legends bang bang* (MLBB).

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