BAB 5 Kesimpulan

5.1 Kesimpulan

Penelitian ini memiliki tujuan untuk memprediksi arah pergerakan suatu saham (naik atau turun) pada hari selanjutnya dengan melakukan pendekatan Eksponensial, *Fourier*, *Gaussian*, Polynomial dan Sin dibandingkan dengan harga aslinya yang dievaluasi menggunakan RMSE dan melakukan validasi data asli dengan prediksi yang sudah dilakukan. Penelitian ini menambahkan metode baru yaitu pendekatan menggunakan sin. Penelitian dilakukan dengan menggunakan data dari tahun 2015 hingga 31 Agustus 2022. Penelitian ini menggunakan sampel data perusahaan LQ45 periode Agustus 2022 hingga Januari 2023. Penelitian ini memiliki hasil sebagai berikut :

1. Metode Sin merupakan metode baru dalam melakukan sebuah prediksi pergerakan harga saham. Dalam penelitian yang ada, Metode Sin lebih baik dibandingkan dengan ke 4 metode yang ada.
2. Penggunaan metode matematika dalam melakukan prediksi pergerakan harga saham dapat dilakukan sebelum harga terbentuk

5.2 Saran

Saran dari penulis untuk penelitian selanjutnya adalah

1. Penelitian menggunakan data yang berbeda beda (1 tahun, 2 tahun dan seterusnya)
2. Penelitian dapat menggunakan aplikasi lain yang dapat mempermudah penghitungan data seperti Excel, Scilab, FreeMat atau aplikasi lainnya.
3. Peneliti selanjutnya dapat memasukan faktor dari ketidakpastian lainnya kedalam perhitungan yang ada

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