



OPERATION ANALYTICS

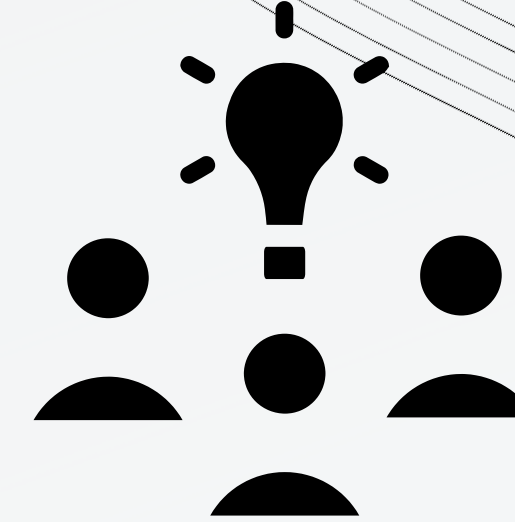
BIKES_HIRED

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LINEAR REGRESSION

- The bike_hires's Multiple R-square shows 54.53% of the variability in the dependent variable can be explained by the model. RMSE 8064.35.
- The F-test is highly significant ($p < 2.2e-16$). Cloud_cover, humidity, pressure, radiation, precipitation, snow_depth, sunshine, mean_temp, and min_temp changed the bicycle hire by 177, 61, 8, 5, 34, 1380, 1380, 32, 1478, and 740 for each increase in one unit. Their p-values are significant, with snow depth and temperature having a large effect on bicycle hire.



```
> lr <- lm(bikes_hired ~ cloud_cover + humidity + pressure + radiation + precipitation + snow_depth + sunshine + mean_temp + min_temp + max_temp, data = bikes_hires)
> summary(lr)
```

```
Call:
lm(formula = bikes_hired ~ cloud_cover + humidity + pressure + radiation + precipitation + snow_depth + sunshine + mean_temp + min_temp + max_temp, data = bikes_hires)
```

Residuals:

Min	1Q	Median	3Q	Max
-29549	-4244	594	4631	34599

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-66522.103	11091.607	-5.998	2.17e-09	***
cloud_cover	177.460	83.654	2.121	0.0339	*
humidity	-61.232	15.691	-3.902	9.67e-05	***
pressure	8.151	1.074	7.590	3.87e-14	***
radiation	5.908	2.865	2.062	0.0393	*
precipitation	-34.199	2.974	-11.501	< 2e-16	***
snow_depth	-1380.500	422.633	-3.266	0.0011	**
sunshine	32.803	7.424	4.418	1.02e-05	***
mean_temp	1478.579	109.698	13.479	< 2e-16	***
min_temp	-740.754	90.921	-8.147	4.81e-16	***
max_temp	56.302	40.784	1.381	0.1675	

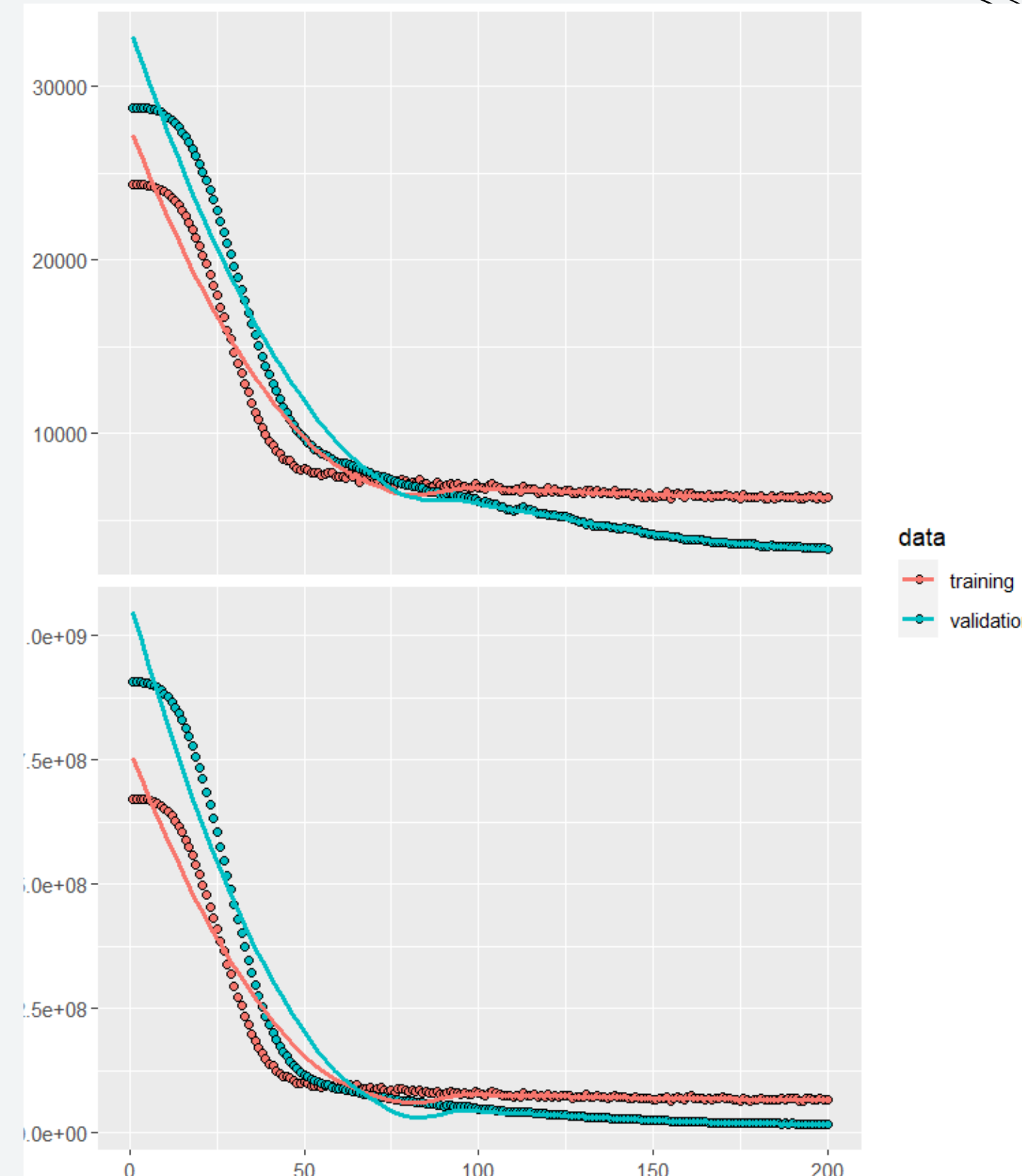
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 6650 on 4374 degrees of freedom
Multiple R-squared: 0.5453, Adjusted R-squared: 0.5443
F-statistic: 524.5 on 10 and 4374 DF, p-value: < 2.2e-16

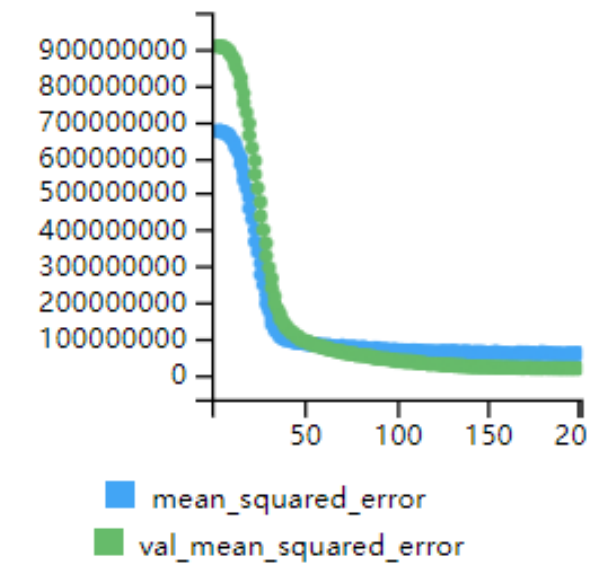
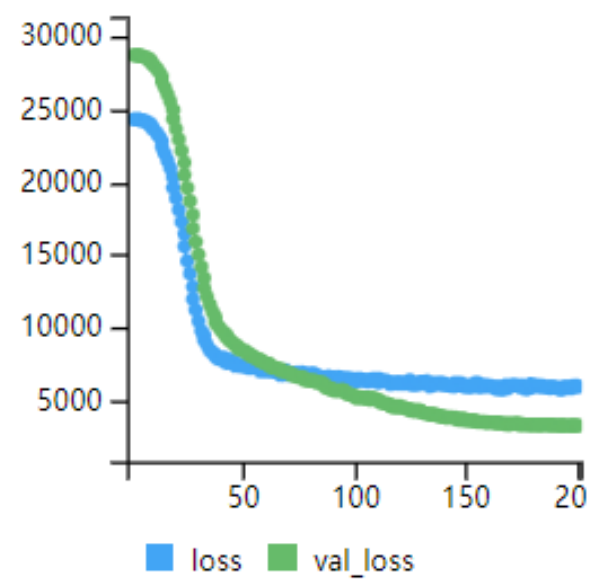
ARTIFICIAL NEURAL NETWORK

The close tracking of the validation loss with the training loss suggests that the model is generalizing well and not overfitting.

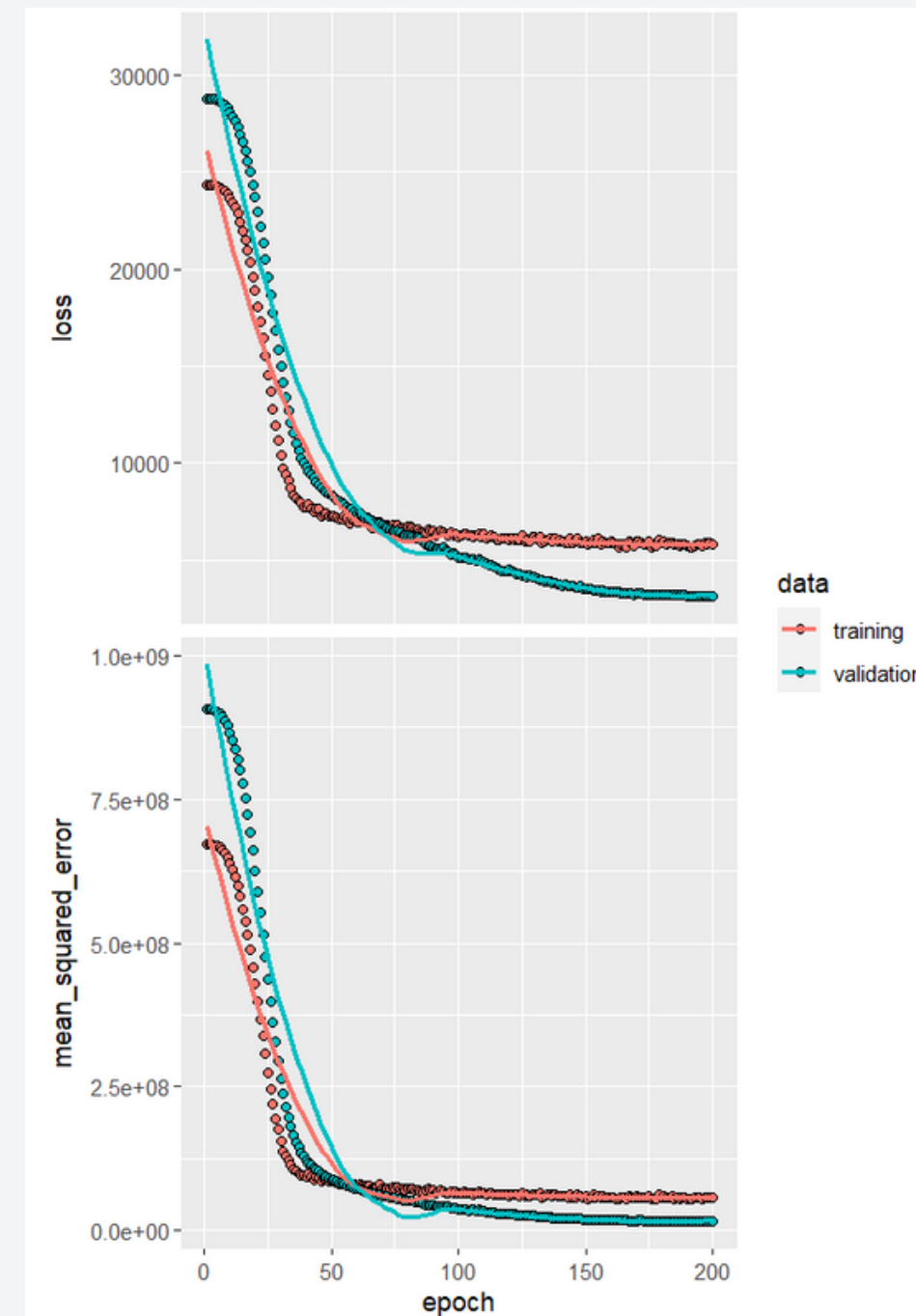
The bike_hires data has a big RMSE which is 8348.106. The data's predictions are off by a large margin on average.



LSTM



The plots represent the loss and MSE from training the LSTM model. There's a sharp decrease in loss and MSE at the start, which indicates initial rapid learning.



CONCLUSION

- *Adjusted R-squared: 54.43%. there is still about 45% of the variability in bike hires that is not explained by linear regression. Time seasonal data also should be considered.*
- *To fulfil the higher demand indicated by the models, bicycle stock should be raised on days with favourable weather, meaning there should be less cloud cover, moderate humidity, and warmer temperatures.*
- *Activities like maintenance and restocking could be planned for days with less favourable weather, when rental numbers are expected to be lower.*



