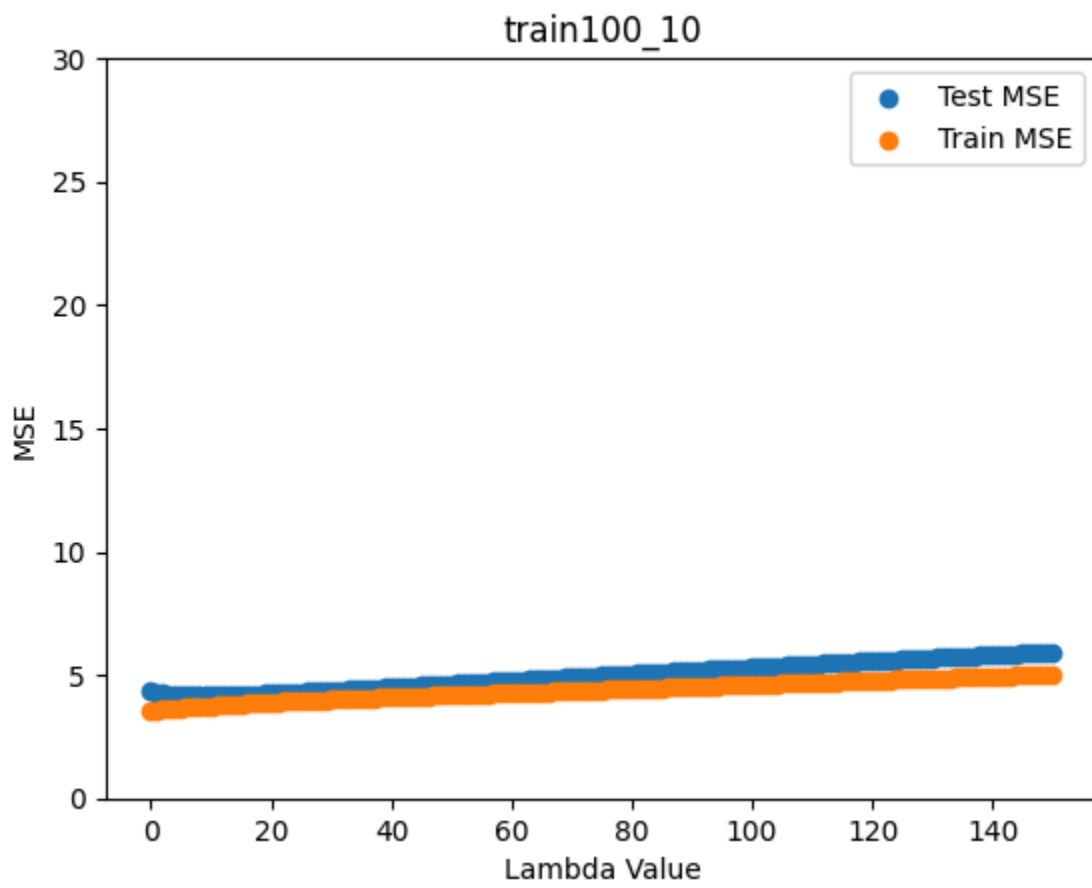
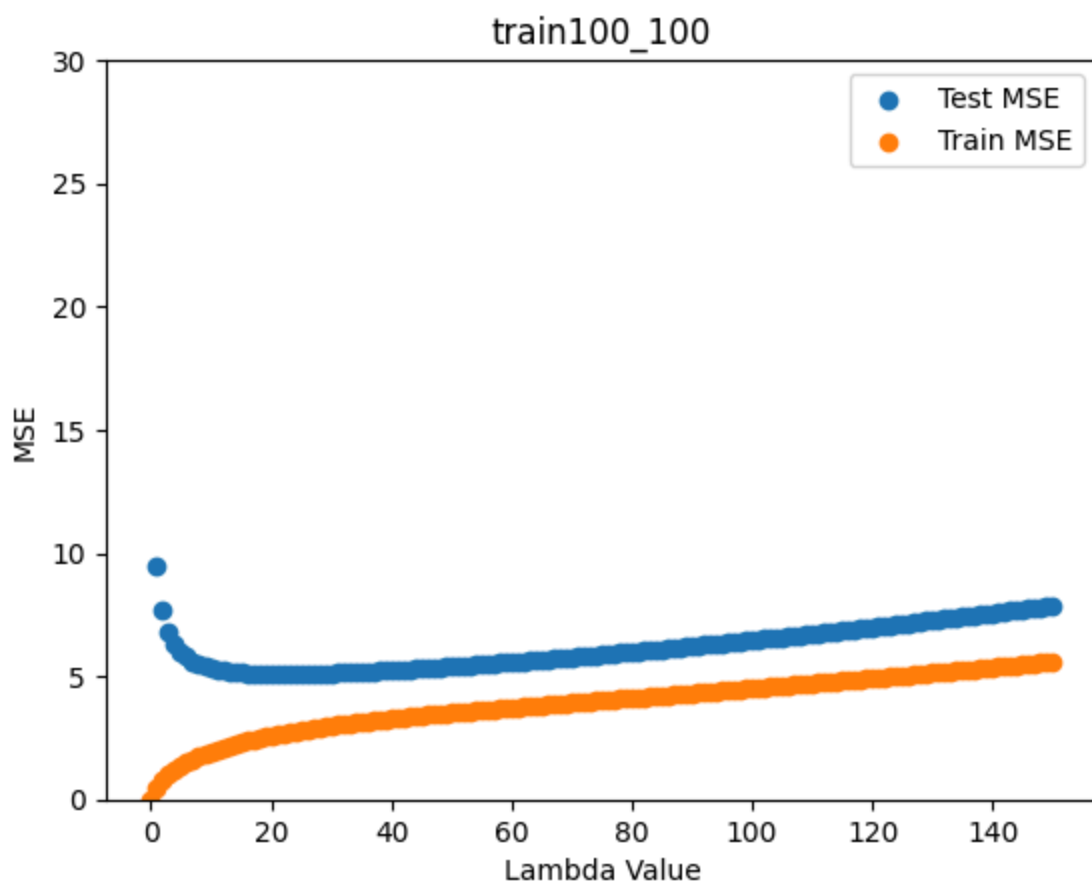
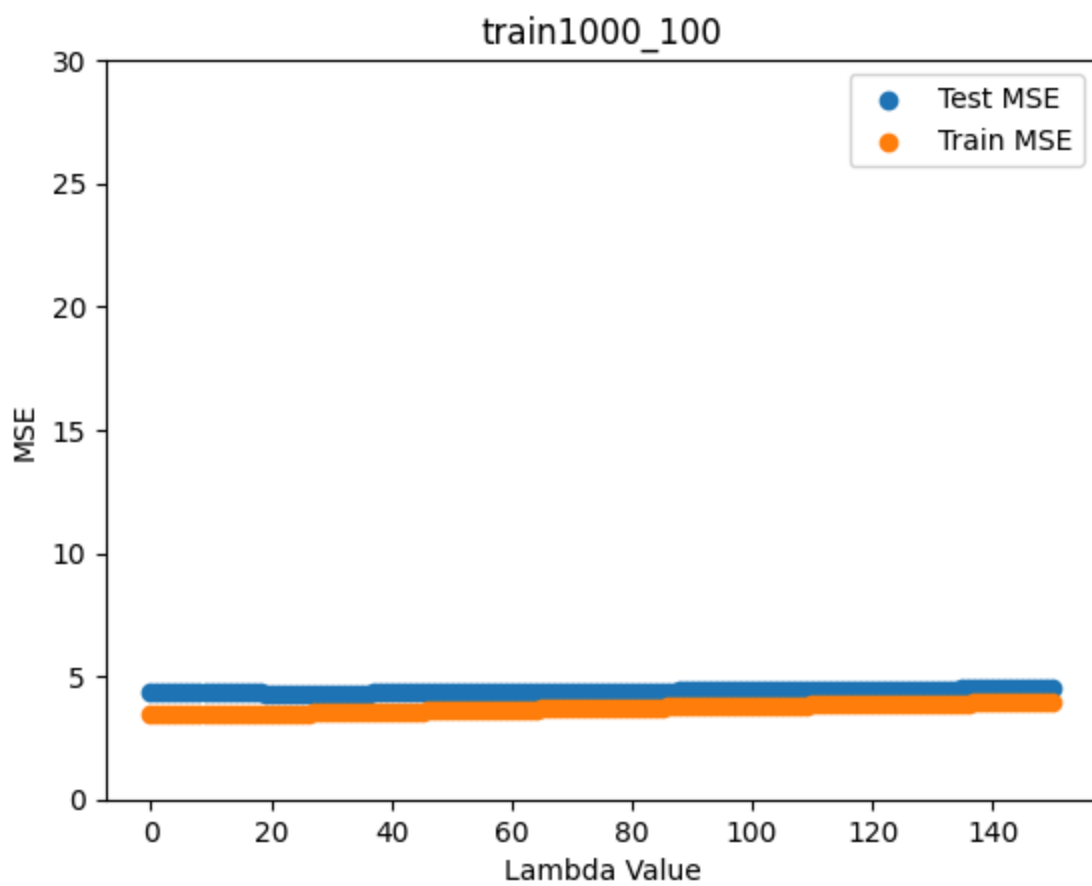
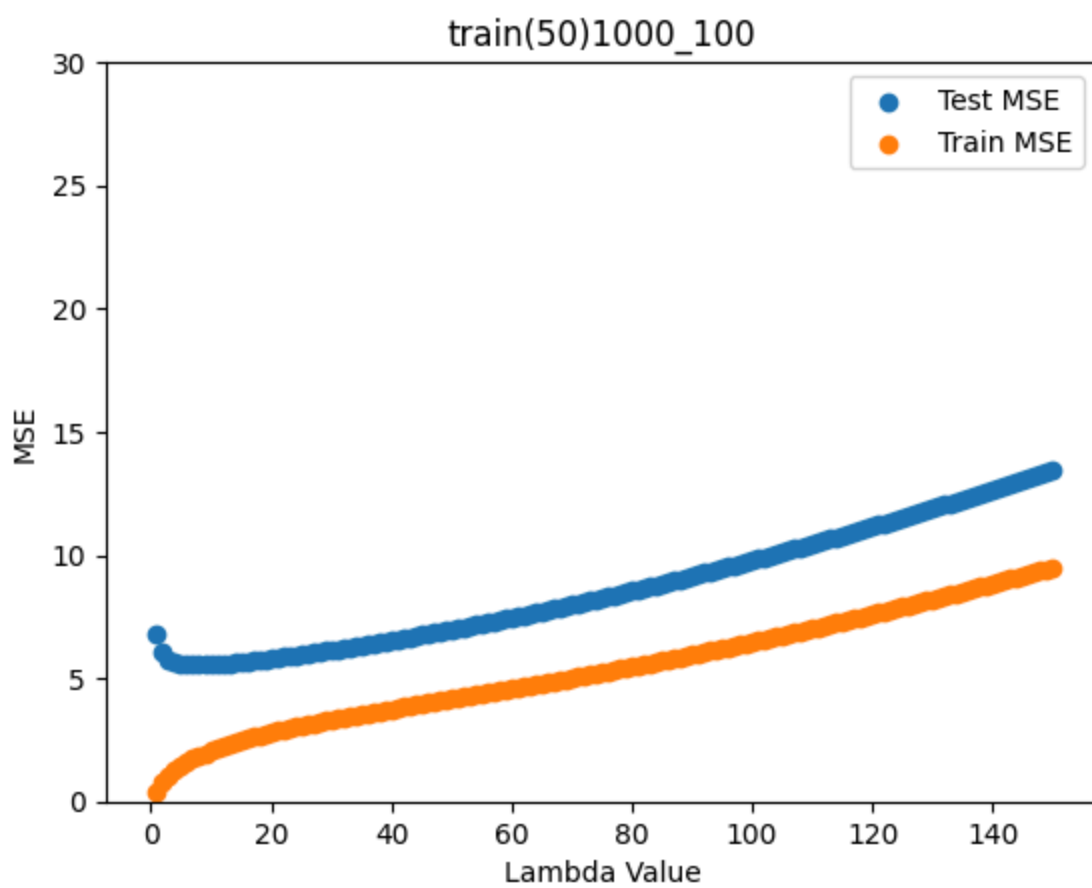


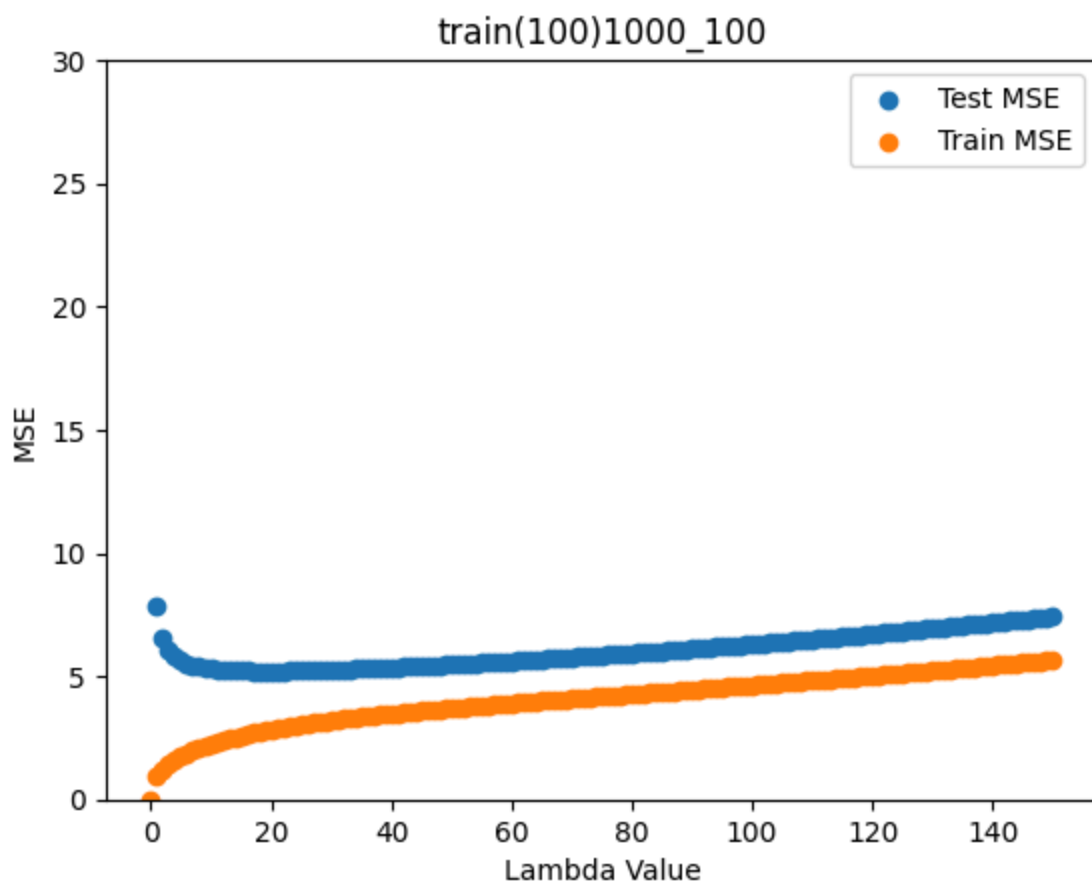
## 1. Graphs

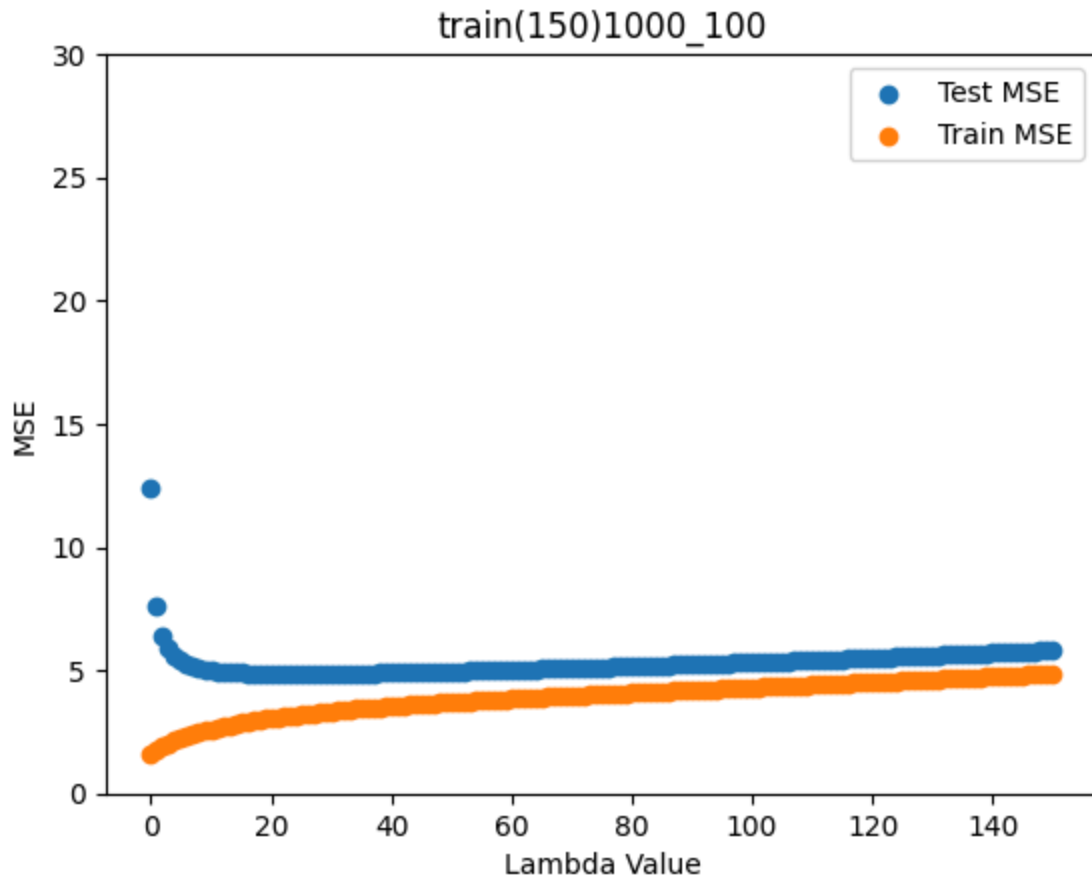












1A.

The values of  $\lambda$  that give the least test set MSE

"train100\_10",  $\lambda = 8$ , MSE = 4.159679

"train100\_100",  $\lambda = 22$ , MSE = 5.0783

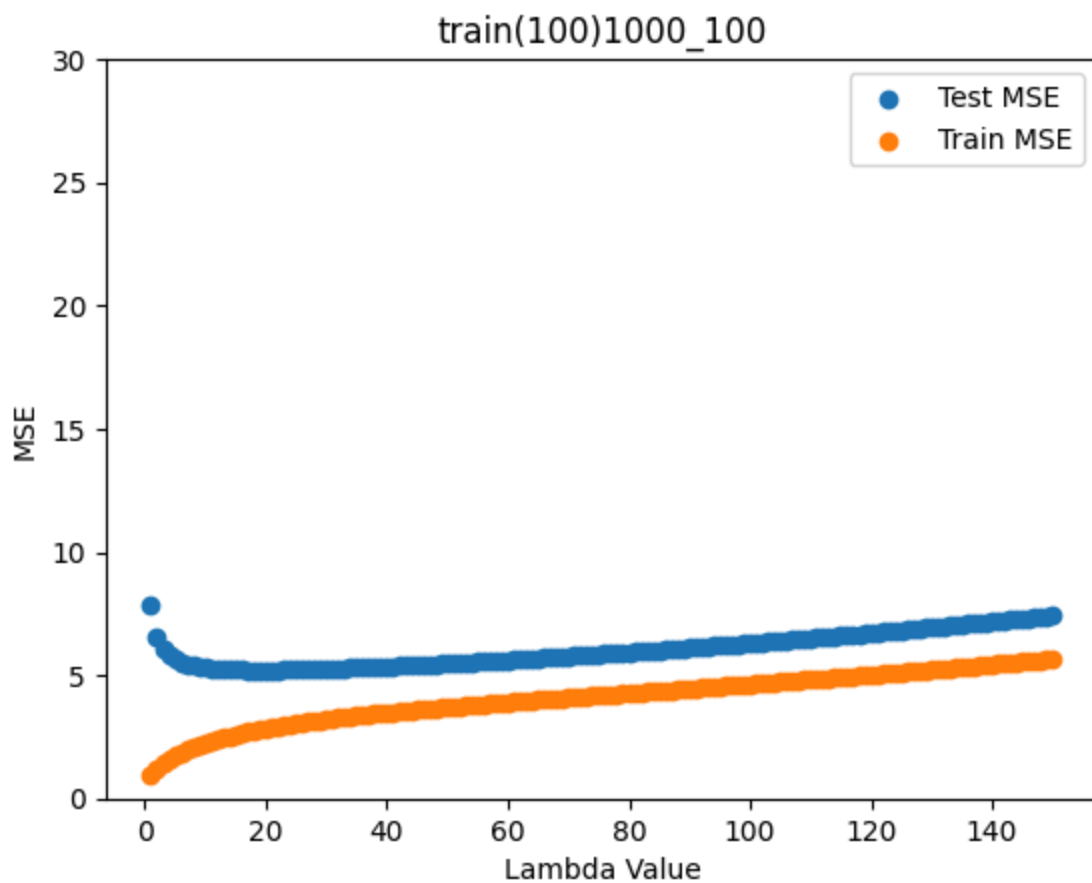
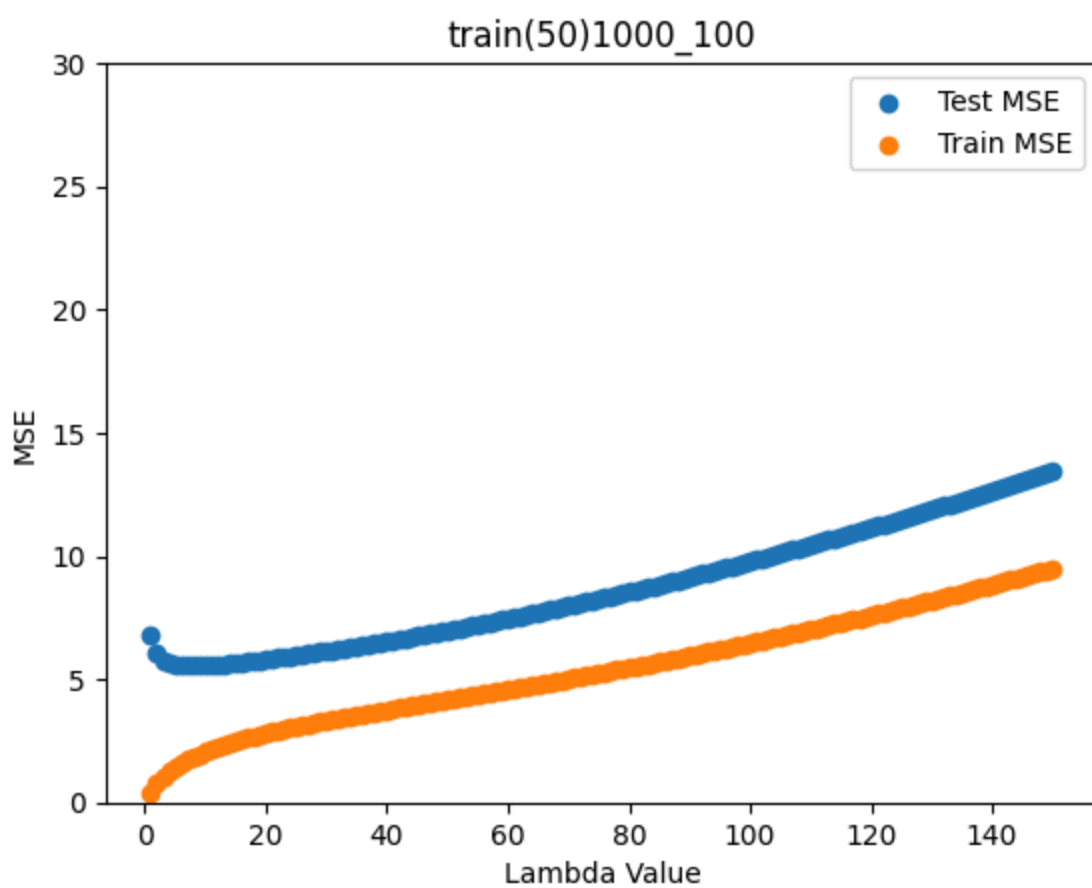
"train1000\_100",  $\lambda = 27$ , MSE = 4.315571

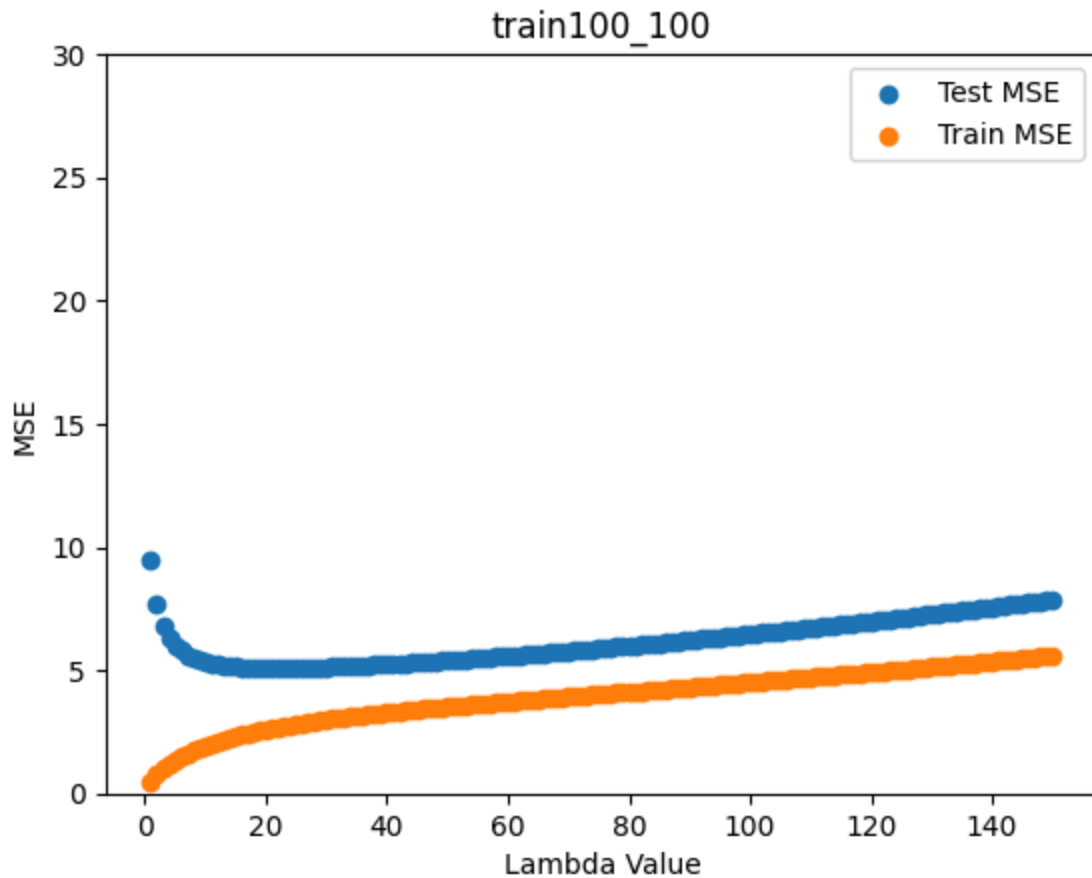
"train(50)1000\_100",  $\lambda = 8$ , MSE = 5.540902

"train(100)1000\_100",  $\lambda = 19$ , MSE = 5.205912

"train(150)1000\_100",  $\lambda = 23$ , MSE = 4.848943

1B. Graphs on next page(s)





**1C.**

The reason why no regularization gives abnormally large MSEs for those datasets is because the size of the datasets is small. With limited amounts of data to train on, testing the data will be more likely to have errors. This is amplified by the lambda value being 0 as there is no regularization being applied within the ridge regression formula. Basically, there are less values and higher variance. Because of these two factors the error will be more extreme because of a lower initial bias.



## 2A.

The values of  $\lambda$  that give the least test set MSE using CV technique

"train100\_10",  $\lambda = 12$ , MSE = 4.158702

"train100\_100",  $\lambda = 20$ , MSE = 4.494045

"train1000\_100",  $\lambda = 39$ , MSE = 4.137141

"train(50)1000\_100",  $\lambda = 24$ , MSE = 5.300696

"train(100)1000\_100",  $\lambda = 30$ , MSE = 4.839469

"train(150)1000\_100",  $\lambda = 46$ , MSE = 4.868577

## 2B.

The lambda values seem larger for the training datasets that are tested on larger testing sets and the MSE values are all smaller if not relatively equal

"train100\_10", pre-cv:  $\lambda = 8$ , MSE = 4.159679

Post-cv:  $\lambda = 12$ , MSE = 4.158702

"train100\_100", pre-cv:  $\lambda = 22$ , MSE = 5.0783

Post-cv:  $\lambda = 20$ , MSE = 4.494045

"train1000\_100", pre-cv:  $\lambda = 27$ , MSE = 4.315571

Post-cv:  $\lambda = 39$ , MSE = 4.137141

"train(50)1000\_100", pre-cv:  $\lambda = 8$ , MSE = 5.540902

Post-cv:  $\lambda = 24$ , MSE = 5.300696

"train(100)1000\_100", pre-cv:  $\lambda = 19$ , MSE = 5.205912

Post-cv:  $\lambda = 30$ , MSE = 4.839469

"train(150)1000\_100", pre-cv:  $\lambda = 23$ , MSE = 4.848943

Post-cv:  $\lambda = 46$ , MSE = 4.868577

**2C.** The drawbacks of cross validation are an increase in computational cost and computational time.

**2D.** The number of folds and number of data points within a fold. The more folds there are the more models there are to train and test. The points there are in a fold the more points you have to train/test a model on.

### 3. Learning Curve Graphs

