

Hyperparameter Optimization: Takeaways



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Syntax

- Using the `expand.grid()` function to automatically create hyperparameter combinations:

```
knn_grid <- expand.grid(k = 1:20)
```

- Plotting to visualize the optimal k value:

```
knn_model <- train(tidy_price ~ accommodates + bathrooms + bedrooms,  
                   data = training_data,  
                   method = "knn",  
                   trControl = train_control,  
                   preProcess = c("center", "scale"),  
                   tuneGrid = knn_grid)  
  
plot(knn_model)
```

Concepts

- Hyperparameters** are parameters that affect the behavior and performance of a model, but are unrelated to the dataset itself.
- Hyperparameter optimization** is the process of finding the optimal hyperparameter value, given a dataset and a machine learning model
- Grid search** is a simple but common hyperparameter optimization technique, which involves evaluating the model performance at different `k` values and selecting the `k` value that results in the lowest validation error. Grid search involves:
 - Selecting a subset of the possible hyperparameter values.
 - Training a model using each of these hyperparameter values.
 - Evaluating each model's performance.
 - Selecting the hyperparameter value that resulted in the lowest error value.

- The general workflow for finding the best model is:
 - Selecting relevant features to use for predicting the target column.
 - Using grid search to find the optimal hyperparameter value for the selected features.
 - Evaluate the model's accuracy and repeat the process.

Resources

- [Difference Between Parameter and Hyperparameter](#)
- [Hyperparameter Optimization](#)
- [caret Documentation](#)



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