# **GA Configuration results**

In this document the results of the GA configuration analysis are included.

Below the results of the models and the respective fitness histories for the GA are shown.

#### The tested parameters are:

- Crossover probability
- Population size
- General mutation probability
- Global mutation probability
- Local mutation probability
- Number of generations

The tested ranges can be found in the respective sections.

# Mean CV Score (measured by R^2) 0.9206506509301235 # Calculate the test metrics test predictions = regressor.predict(X test\_rmse = np.sqrt(mean\_squared\_error test r2 = r2 score(y test, test predict print("Test RMSE:", test\_rmse) print("Test NRMSE score:", test\_rmse / print("Test R2 Score:", test\_r2) print(f"Tuning completed in: {end time ✓ 0.0s Test RMSE: 22067.814298356974 Test R2 Score: 0.9263404479317279 Tuning completed in: 154.897918 seconds.

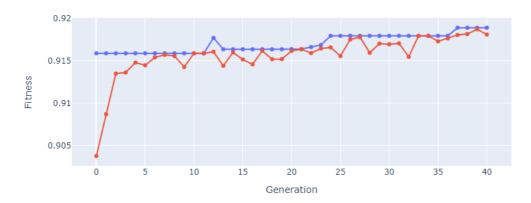
# Crossover probability: 0.1

#### Crossover probability: 0.5

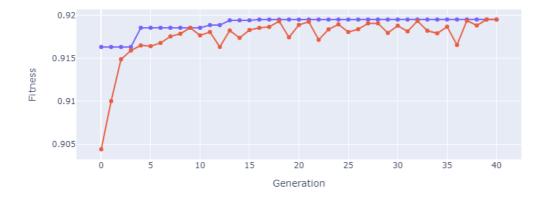
# Crossover probability: 0.3

#### Crossover probability: 0.7

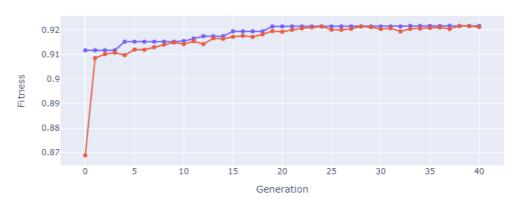
Fitness history per generation - Crossover probability = 0.1



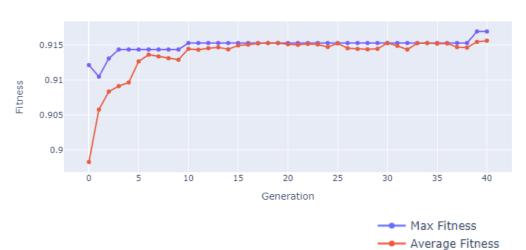
Fitness history per generation - Crossover probability = 0.5



Fitness history per generation - Crossover probability = 0.3



Fitness history per generation - Crossover probability = 0.7



# Population size: 5

# Population size: 20

# Population size: 15

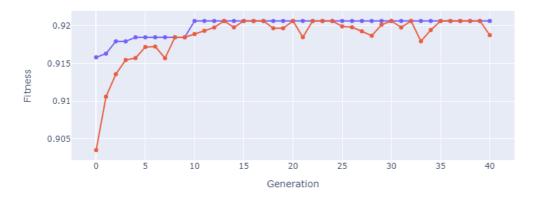
#### Population size: 25

```
# Calculate the test metrics
test_predictions = regressor.predict(X_testest_rmse = np.sqrt(mean_squared_error(y_ttest_r2 = r2_score(y_test, test_prediction)

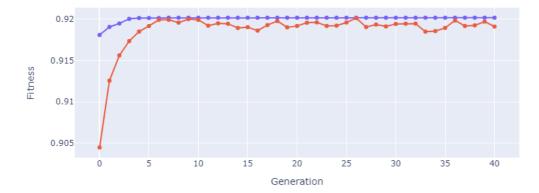
print("Test RMSE:", test_rmse)
print("Test NRMSE score:", test_rmse / (y_print("Test R2 Score:", test_r2)
print(f"Tuning completed in: {end_time-statestary o.1s

Test RMSE: 19982.085779988778
Test NRMSE score: 0.0269223789191871
Test R2 Score: 0.9396062398122474
Tuning completed in: 1313.378993 seconds.
```

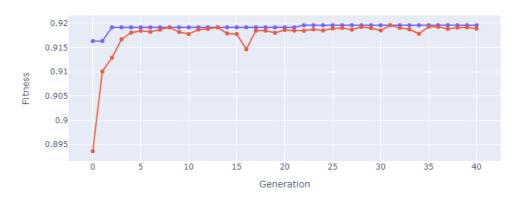
Fitness history per generation - Population size = 5



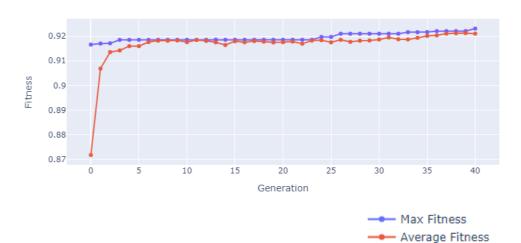
Fitness history per generation - Population size = 20



Fitness history per generation - Population size = 15



Fitness history per generation - Population size = 25



#### General mutation probability: 0.1

# General mutation probability: 0.5

# General mutation probability: 0.3

```
# Calculate the test metrics
test_predictions = regressor.predict(X
test_rmse = np.sqrt(mean_squared_error
test_r2 = r2_score(y_test, test_predic

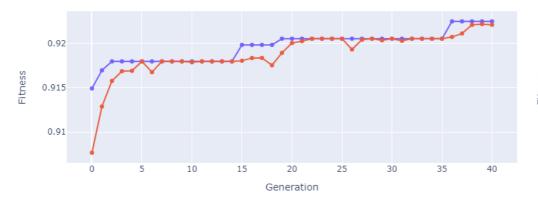
print("Test RMSE:", test_rmse)
print("Test NRMSE score:", test_rmse /
print("Test R2 Score:", test_r2)
print(f"Tuning completed in: {end_time

✓ 0.0s

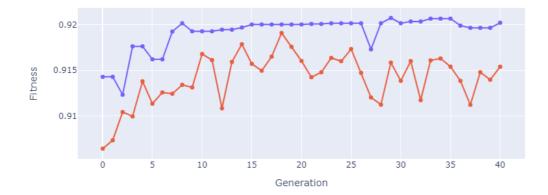
Test RMSE: 20590.252531095866
Test NRMSE score: 0.02774177764961159
Test R2 Score: 0.9358740549795291
Tuning completed in: 231.000848 seconds.
```

# General mutation probability: 0.7

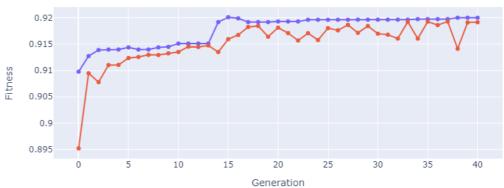
Fitness history per generation - Mutation probability = 0.1



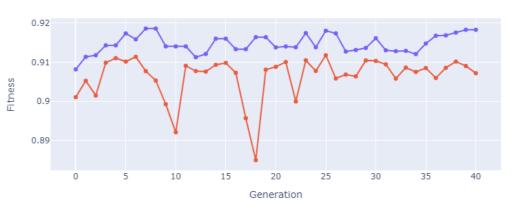
Fitness history per generation - Mutation probability = 0.5



Fitness history per generation - Mutation probability = 0.3



Fitness history per generation - Mutation probability = 0.7



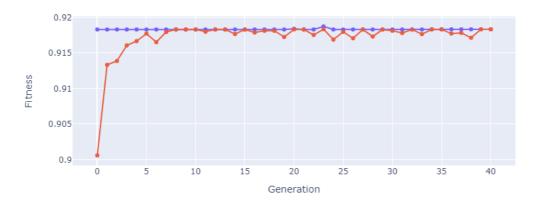
#### Global mutation probability: 0.1

#### Global mutation probability: 0.5

# Global mutation probability: 0.3

#### Global mutation probability: 0.7

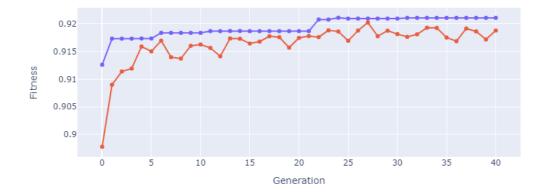
Fitness history per generation - Global mutation probability = 0.1



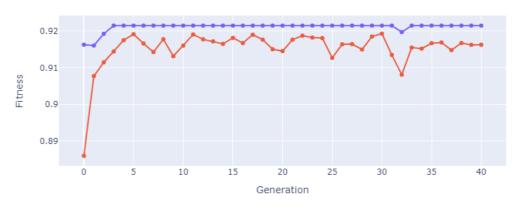
Fitness history per generation - Global mutation probability = 0.3



Fitness history per generation - Global mutation probability = 0.5



Fitness history per generation - Global mutation probability = 0.7



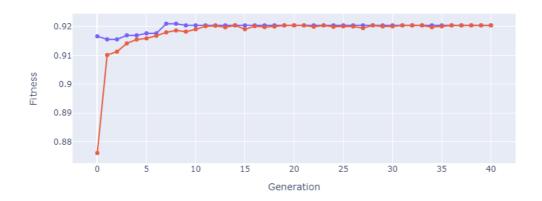
#### Local mutation probability: 0.1

# Local mutation probability: 0.5

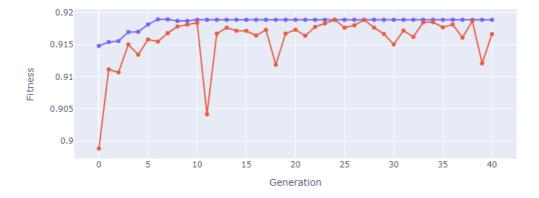
# Local mutation probability: 0.3

# Local mutation probability: 0.7

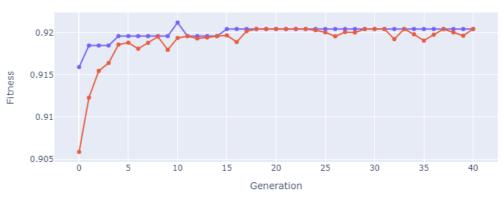
Fitness history per generation - Local mutation probability = 0.1



Fitness history per generation - Local mutation probability = 0.5



Fitness history per generation - Local mutation probability = 0.3



Fitness history per generation - Local mutation probability = 0.7



# Number of generations: 10

```
# Calculate the test metrics
test_predictions = regressor.predict()
test_rmse = np.sqrt(mean_squared_error
test_r2 = r2_score(y_test, test_predict)

print("Test RMSE:", test_rmse)
print("Test NRMSE score:", test_rmse)
print("Test R2 Score:", test_rmse)
print("Test R2 Score:", test_rase)
print(f"Tuning completed in: {end_timest R2 Score: 0.027464814460937115}
Test R2 Score: 0.027464814460937115
Test R2 Score: 0.9371480806975142
Tuning completed in: 133.334631 seconds.
```

# Number of generations: 30

# Number of generations: 20

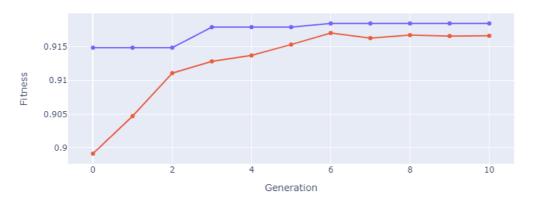
```
# Calculate the test metrics
test_predictions = regressor.predict()
test_rmse = np.sqrt(mean_squared_error
test_r2 = r2_score(y_test, test_predic

print("Test RMSE:", test_rmse)
print("Test NRMSE score:", test_rmse /
print("Test R2 Score:", test_r2)
print(f"Tuning completed in: {end_time
✓ 0.0s

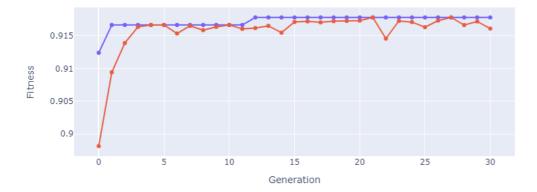
Test RMSE: 20351.344947789778
Test NRMSE score: 0.027419891308253015
Test R2 Score: 0.9373535215331206
Tuning completed in: 105.416116 seconds.
```

#### Number of generations: 50

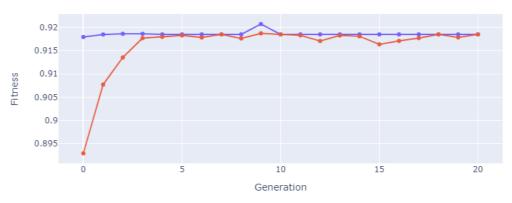
Fitness history per generation - Number of generations = 10



Fitness history per generation - Number of generations = 30



Fitness history per generation - Number of generations = 20



Fitness history per generation - Number of generations = 50

