```
# IMPORTANT: SOME KAGGLE DATA SOURCES ARE PRIVATE
# RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES.
import kagglehub
kagglehub.login()
# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.
harshanrs_matrix_path = kagglehub.dataset_download('harshanrs/matrix')
print('Data source import complete.')
#CS547 - ADVANCED TOPICS IN SOFTWARE ENGINEERING
#ASSIGNMENT 1 - Part 2
#Group: CS547Assignment1Part2Group47
#Members:
#1. HARSHAN RETHINAVELU SELVAKUMAR - 202480548
#2. MANOJ KUMAR DHARMARAJ - 202468855
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
       print(os.path.join(dirname, filename))
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Sa
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session
/kaggle/input/matrix/bigfaultmatrixplustime.txt
     /kaggle/input/matrix/smallfaultmatrixplustime.txt
#Importing the necessary libraries
import random
import matplotlib.pyplot as plt
from deap import creator, base, tools, algorithms
#Function that loads the data of the file including the execution time
def load_file(filepath):
    #Create an empty list
    tests=[]
    with open(filepath, 'r') as file:
        for line in file:
           #Taking each line of the file seperately, removing any leading whitespaces
           #Splitting the line into list of strings using commas
           seperate=line.strip().split(',')
           test_num=seperate[0]
           #Faults upto the last element
           fault=list(map(int,seperate[1:-1]))
           #Execution time(last element)
           execution_time=float(seperate[-1])
           #Adding the test id, fault and execution time
           tests.append((test_num,fault,execution_time))
    return tests
```

```
def fit(individual,tests):
    #Creating an empty list to store the selected tests
    sel_tests=[]
    for i in range(len(individual)):
        if individual[i]==1:
            sel_tests.append(tests[i])
    tot execution time=0
    #Tracking unique faults detected
    detected_faults=set()
    #Looping through each selected test
    for _,faults,execution_time in sel_tests:
        tot_execution_time+=execution_time
        #Update the list (Unique faults)
        for index,fault in enumerate(faults):
            if fault==1:
                detected_faults.add(index)
    #Total number of faults detected
    tot_detected_faults=len(detected_faults)
    #Return the objectives
    return tot_detected_faults,tot_execution_time
#Creating DEAP classes
creator.create("FitnessMulti",base.Fitness,weights=(1.0,-1.0))
\verb|creator.create| ("Individual", list, fitness = \verb|creator.FitnessMulti|)| \\
#Defining DEAP toolbox
def tbox(tests):
   toolbox=base.Toolbox()
    toolbox.register("attr_bool",random.randint,0,1)
    toolbox.register("individual", tools.initRepeat, creator.Individual, toolbox.attr\_bool, n=len(tests))
    toolbox.register("population",tools.initRepeat,list,toolbox.individual)
    toolbox.register("evaluate",fit,tests=tests)
    #Defining the genetic operators
    toolbox.register("mate",tools.cxUniform,indpb=0.5)
    toolbox.register("mutate",tools.mutFlipBit,indpb=0.2)
    toolbox.register("select",tools.selNSGA2)
    return toolbox
#Genetic Algorithm
def ga(toolbox,tests,pop_size=50,gen=25,cxpb=0.7,mutpb=0.3):
    #Initializing the population
    pop=toolbox.population(n=pop_size)
    #Evaluating the initial population
    for ind in pop:
       ind.fitness.values=toolbox.evaluate(ind)
    algorithms.eaMuPlusLambda(pop,toolbox,mu=pop_size,lambda_=pop_size,
                              cxpb=cxpb,mutpb=mutpb,ngen=gen,verbose=False)
    #Pareto Front
    pareto=tools.sortNondominated(pop,len(pop),first_front_only=True)[0]
    print(f"\nResults for GA (pop_size={pop_size},gen={gen},cxpb={cxpb},mutpb={mutpb}):")
    print(f"\nTotal final solutions from pareto front:{len(pareto)}")
    #Printing the best individuals
    print("Best Individuals in pareto front:")
    for i,individual in enumerate(pareto,start=1):
        sel_tests=[tests[index][0] for index,bit in enumerate(individual) if bit==1]
        print(f"\tIndividual{i}:{sel_tests}")
        print(f"\tIndividual{i} Fitness: \{individual.fitness.values\}\n")
    return pareto
```

```
def cal_metrics(pareto):
    #Average of faults detected
    faults_avg=np.mean([ind.fitness.values[0] for ind in pareto])
    #Average execution time
    time_avg=np.mean([ind.fitness.values[1] for ind in pareto])
    #Max of faults detected
    faults max=max([ind.fitness.values[0] for ind in pareto])
    #Minimum execution time
    time_min=min([ind.fitness.values[1] for ind in pareto])
    return{"Average of faults detected": faults_avg,
          "Average execution time": time_avg,
          "Max of faults detected": faults_max,
          "Minimum Execution time": time min}
#Random Search
from deap.creator import Individual
def random_search(tests,runs=10):
    #Creating a list to store the results
    result=[]
    for _ in range(runs):
        #Random Individual
        individual=[random.randint(0,1) for _ in range(len(tests))]
        #Evaluating the fitness
        faults,time=fit(individual,tests)
        #Creating DEAP individual
        ind=Individual(individual)
        ind.fitness.values=(faults,time)
        result.append(ind)
    #Pareto front
    pareto=tools.sortNondominated([sol for sol in result],len(result),first_front_only=True)[0]
    return pareto, result
#Visualization
def visualize(ga_pareto,random_pareto):
    fault_GA=[ind.fitness.values[0] for ind in ga_pareto]
    time_GA=[ind.fitness.values[1] for ind in ga_pareto]
    fault_random=[sol[0] for sol in random_pareto]
    time_random=[sol[1] for sol in random_pareto]
    plt.figure(figsize=(10,6))
   plt.scatter(time_random,fault_random,label="Random Search",alpha=0.6)
    plt.scatter(time_GA,fault_GA,label="Genetic Algorithm",marker='o',color='red')
   plt.xlabel("Execution Time")
   plt.ylabel("Detected Faults")
    plt.title("Gentic Algorithm vs Random Search")
   plt.legend()
   plt.grid()
   plt.show()
#Main funtion
if __name__=="__main__":
    #Reading the datasets
    test1=load_file("/kaggle/input/matrix/smallfaultmatrixplustime.txt")
    test2=load_file("/kaggle/input/matrix/bigfaultmatrixplustime.txt")
    #Parameters sets for diverse solutions
    parameters=[{"pop_size":50,"gen":25,"cxpb":0.7,"mutpb":0.3},
               {"pop_size":50,"gen":25,"cxpb":0.6,"mutpb":0.4},
               {"pop_size":50,"gen":25,"cxpb":0.8,"mutpb":0.2},]
```

#Function to compute the metrics from the pareto front

```
#Using for loop to go through the datasets and parameters
for dataset,tests in [("smallfaultmatrixplustime",test1),("bigfaultmatrixplustime",test2)]:
   print(f"\n{dataset}:")
   #Creating toolbox
   toolbox=tbox(tests)
    for parameter in parameters:
        #Genetic Algorithm...
        pareto=ga(toolbox,tests,pop_size=parameter["pop_size"],
                  gen=parameter["gen"],cxpb=parameter["cxpb"],mutpb=parameter["mutpb"])
        \hbox{\#Computing metrics for $GA$}
        ga_metrics=cal_metrics(pareto)
        print(f"\nGA Metrics:" )
       print(ga_metrics)
        #Random Search...
       pareto_random,result=random_search(tests,runs=10)
        #Visualizing the results
        visualize(pareto,pareto_random)
```

smallfaultmatrixplustime:

Results for GA (pop size=50,gen=25,cxpb=0.7,mutpb=0.3):

Total final solutions from pareto front:3

Best Individuals in pareto front:

Individual1:['t0', 't6', 't14', 't19', 't24', 't33', 't36', 't44', 't52', 't56', 't59', 't65', 't67', 't72', 't77', 't88', Individual1 Fitness:(8.0, 193.0)

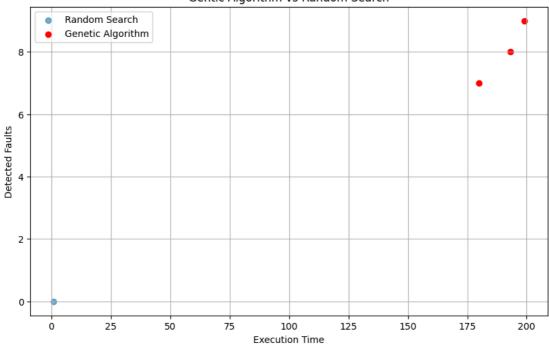
Individual2:['t0', 't1', 't6', 't14', 't16', 't20', 't24', 't32', 't35', 't36', 't44', 't52', 't59', 't73', 't77', 't88', '1 Individual2 Fitness:(7.0, 180.0)

Individual3:['t0', 't1', 't4', 't6', 't14', 't16', 't20', 't24', 't36', 't40', 't44', 't45', 't52', 't55', 't56', 't59', 't6 Individual3 Fitness:(9.0, 199.0)

GA Metrics:

{'Average of faults detected': 8.0, 'Average execution time': 190.666666666666, 'Max of faults detected': 9.0, 'Minimum Execution

Gentic Algorithm vs Random Search



Results for GA (pop_size=50,gen=25,cxpb=0.6,mutpb=0.4):

Total final solutions from pareto front:2

Best Individuals in pareto front:

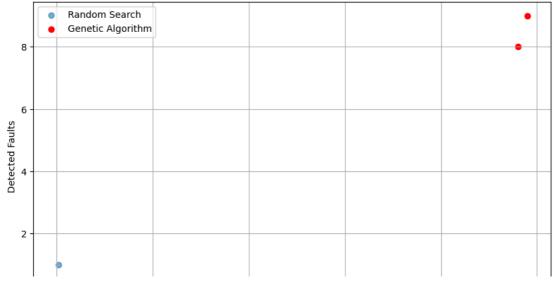
Individual1:['t0', 't3', 't4', 't11', 't12', 't15', 't16', 't27', 't35', 't36', 't37', 't42', 't43', 't44', 't45', 't48', 't Individual1 Fitness:(9.0, 245.0)

Individual2:['t0', 't1', 't12', 't30', 't35', 't36', 't37', 't42', 't44', 't45', 't49', 't50', 't52', 't54', 't57', 't62', Individual2 Fitness:(8.0, 240.0)

GA Metrics:

{'Average of faults detected': 8.5, 'Average execution time': 242.5, 'Max of faults detected': 9.0, 'Minimum Execution time': 240.0]

Gentic Algorithm vs Random Search

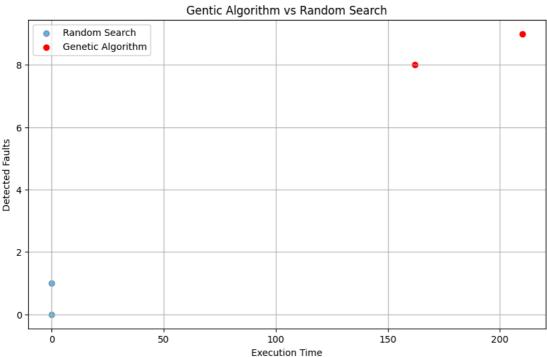




Results for GA (pop_size=50,gen=25,cxpb=0.8,mutpb=0.2):

GA Metrics:

{'Average of faults detected': 8.5, 'Average execution time': 186.0, 'Max of faults detected': 9.0, 'Minimum Execution time': 162.0]



bigfaultmatrixplustime:

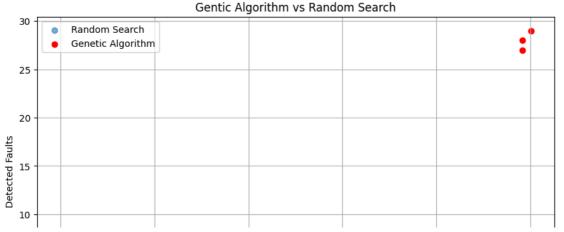
Results for GA (pop_size=50,gen=25,cxpb=0.7,mutpb=0.3):

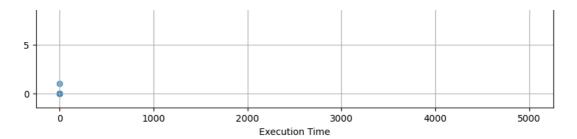
Individual2 Fitness:(28.0, 4916.0)

Individual3:['t100', 't1000', 't10001', 't10002', 't10003', 't10007', 't10009', 't10010', 't10014', 't10015', 't10016', '

GA Metrics:

{'Average of faults detected': 28.0, 'Average execution time': 4945.66666666667, 'Max of faults detected': 29.0, 'Minimum Execution





Results for GA (pop_size=50,gen=25,cxpb=0.6,mutpb=0.4):

Total final solutions from pareto front:3

Best Individuals in pareto front:

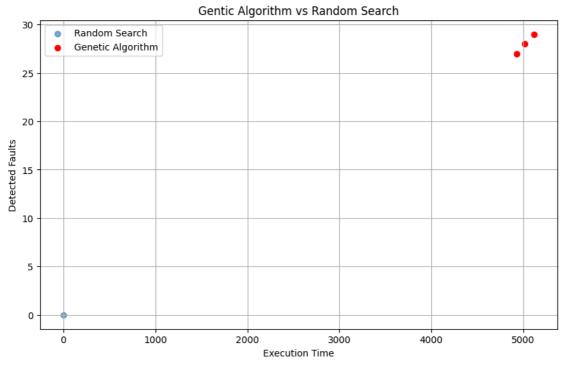
Individual1:['t100', 't10000', 't10001', 't10004', 't10007', 't10009', 't10012', 't10013', 't10014', 't10015', 't10021', 't1
Individual1 Fitness:(29.0, 5115.0)

Individual2:['t10001', 't10002', 't10003', 't10006', 't10007', 't10009', 't1001', 't10011', 't10013', 't10014', 't10016', 't
Individual2 Fitness:(28.0, 5015.0)

Individual3:['t10001', 't10002', 't10003', 't10005', 't10006', 't1001', 't10013', 't10014', 't10016', 't1002', 't10020', 't1
Individual3 Fitness:(27.0, 4927.0)

GA Metrics:

{'Average of faults detected': 28.0, 'Average execution time': 5019.0, 'Max of faults detected': 29.0, 'Minimum Execution time': 492



Results for GA (pop_size=50,gen=25,cxpb=0.8,mutpb=0.2):

Total final solutions from pareto front:1

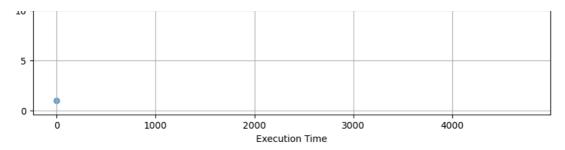
 ${\tt Best\ Individuals\ in\ pareto\ front:}$

Individual1:['t1000', 't10003', 't10007', 't10009', 't10011', 't10016', 't10017', 't10022', 't10029', 't10032', 't10036', 't
Individual1 Fitness:(29.0, 4752.0)

GA Metrics:

{'Average of faults detected': 29.0, 'Average execution time': 4752.0, 'Max of faults detected': 29.0, 'Minimum Execution time': 475

Gentic Algorithm vs Random Search Random Search Genetic Algorithm 25 20 10



#References

#https://github.com/DEAP/deap

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