







Project Progress Review #3

(High Level / Low-Level Design)

Project Title

: Efficient Python Genetic Algorithm Framework

Project ID

: PW19CGM01

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Project Team

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Project Abstract and Scope

Our Genetic Algorithm Framework is proposed to be a very efficient, generic Framework where users can easily simulate all variations of Genetic Algorithms very easily. It is a usable way to **explore the problem solving ability of Genetic Algorithms**.

Usage of Spark to explore improvements in GA by using large populations

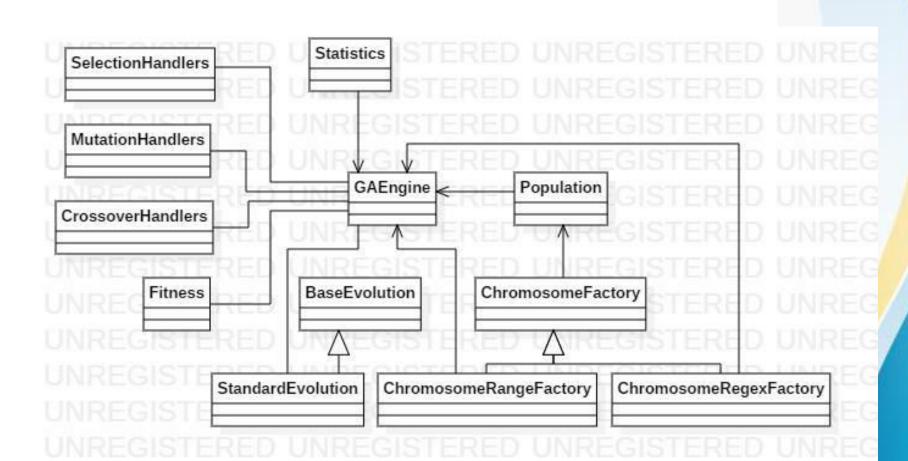


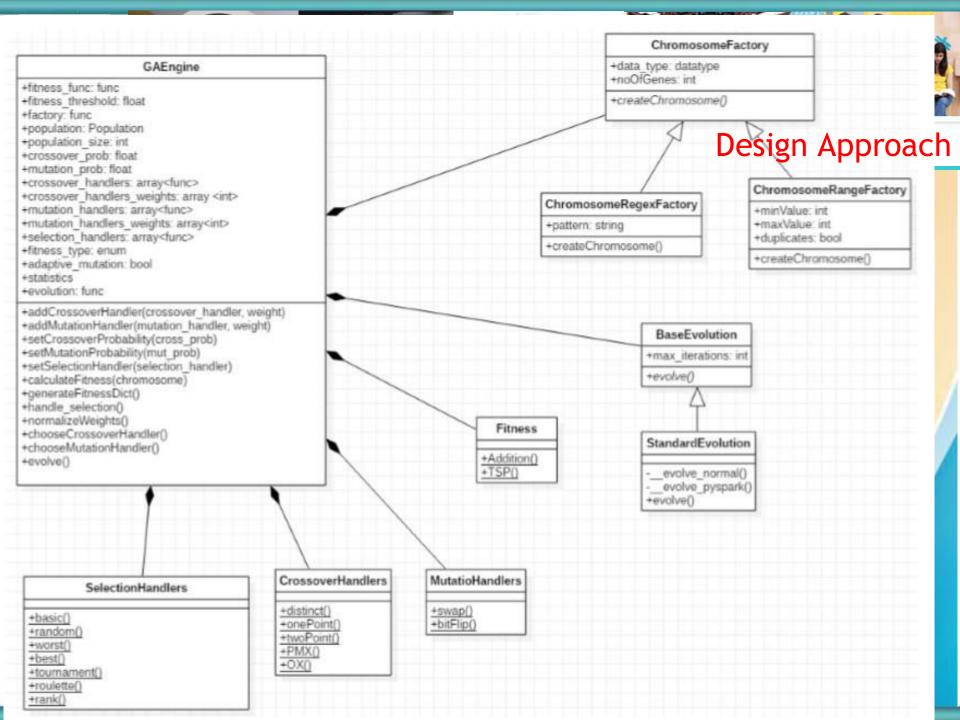






Design Approach













Design Approach

Benefits of this approach

- Follows good object oriented API principles of SOLID
- Is very generic in nature (users can define custom evolution style, crossovers, mutations, selection, chromosome factories easily)

Are there any drawbacks?

- Takes more lines of code than a less generic API.
- Very high dependence on good documentation

Alternate design approaches, if any.

High Level, less generic API with all the basic types of operations so as to reduce the number of lines of code needed









Design Constraints, Assumptions & Dependencies

Design Constraints:

- Need for spark to be correctly installed and configured
- Since it is generic in nature, need for more lines of code

Assumptions:

- Fitness type can be max, min, equal
- Crossover, mutation operations are expensive compared to generation of indexes RDD and spark operations

Dependencies:

Pyspark: Need for Apache Spark, Scala, JVM, pyspark installation

for parallel execution using spark to work

Matplotlib: Needed for graph generations







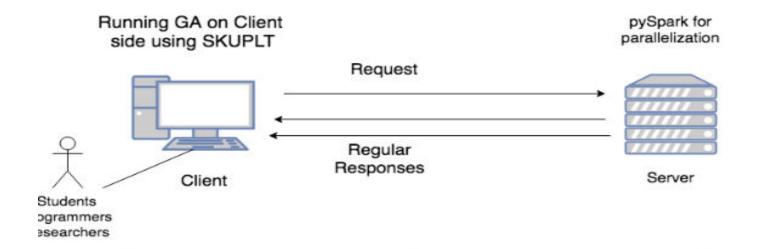


Design Description

Step 1: Take users inputs about the GA from the UI Step 2: Convert the user inputs into python code which uses our

GA API

Step 3: Run the code and observe GA execution and results



Choice:User can choose the run the code on the client side or the server side.

Client:no need to contact server for every GA generation update, no db support, no parallelization

Server:need for regular updates, db support, parallelization pySpark,
"Genetic Algorithm as a Service"









Design Description

GA Online Simulation			
Gene number of genes / chromosome			
crossover		crossover type	
mutation prob		mutation type	
population size		selection	
fitness			
MAX ITER			
	SIMULATE ON CLIENT	SIMULATE ON SERVER	



Show analytics, evolution details on simulation



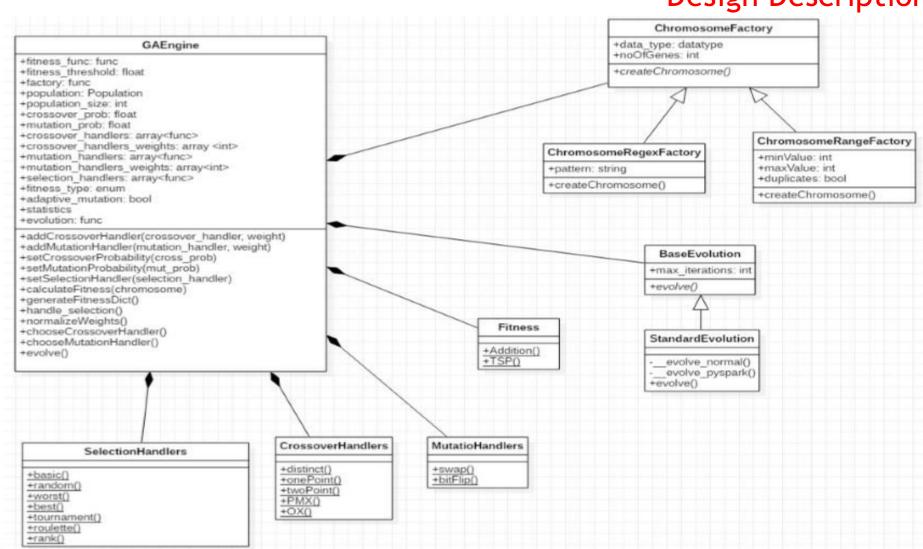








Design Description





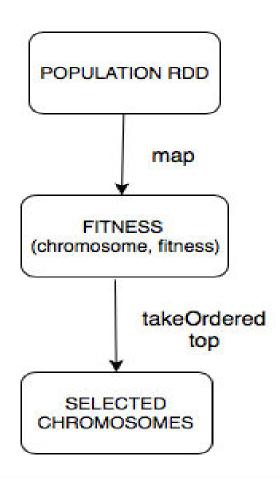






Spark DAG Design Description: Selection

SELECTION DAG



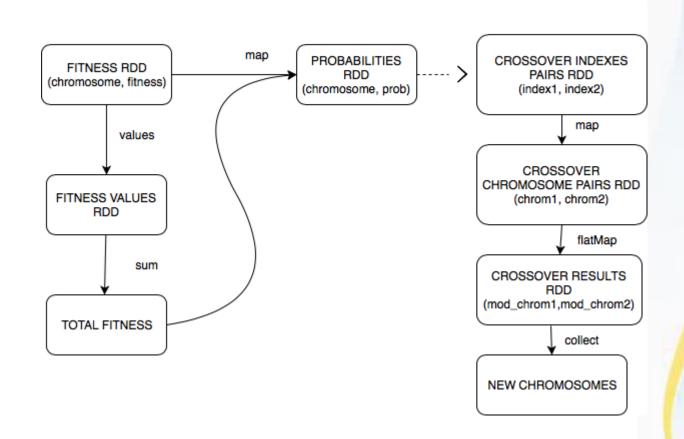






Spark DAG Design Description: Crossover

CROSSOVER DAG





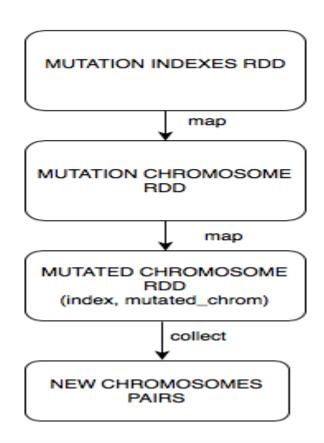






Spark DAG Design Description: Mutation

MUTATION DAG





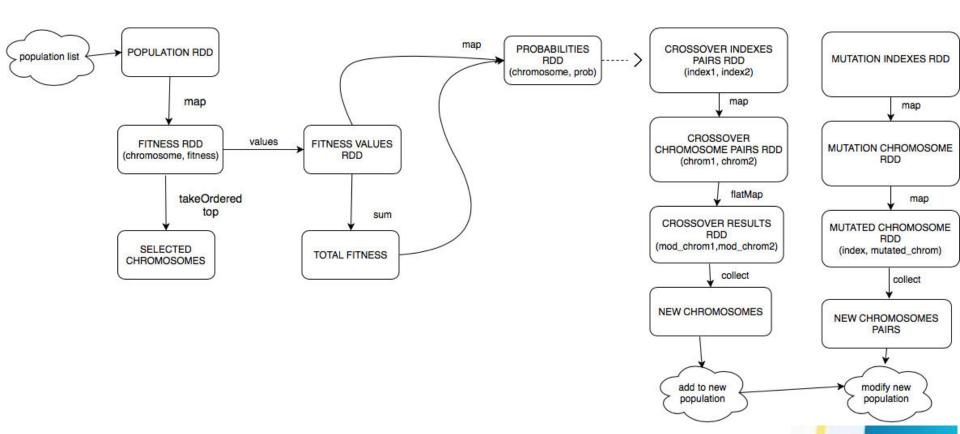






Spark DAG Design Description: Overall

OVERALL GA DAG











Technologies Used

- Python 3.0 and above
- Flask for running algo as a service on server which receives parameters from UI.
- Pyspark for parallelisation of operations in algo
- Pytest for testing purposes.
- Matplotlib For statistic graph plotting









Divided work into three branches of activities

- 1. Implementation of classes
- 2. Website UI and functionality implementation
- 3. Documentation









1. Implementation of API

- Completed major parts of implementation of Regex/Range ChromosomeFactory, Utils, GAEngine, Statistics, Evolution classes
- Statistics of max, min, avg fitness
- Implementation of adaptive mutation GA optimization to change mutation rate based on diversity
- Implementation of a basic version of evolution using pySpark run on a single machine.









- 2. Website UI and functionality implementation
- Designed the basic frontend webpages using HTML, CSS, JS, jQuery.
- Designed basic Flask backend to serve webpages
- Done with conversion of user inputs into equivalent python code implementation
- Started with basic polling from website to get each generation data one by one









3. Documentation

- Worked on commenting every python file made to aid users in understanding our API
- Started on working on official pyGenetic documentation which is hosted at readthedocs.org









Work Left:

- Exploring GA on pySpark: Solving famous problems faster
- Implementation of more optimizations
- Implementation of other ML using GAs
- Website functionalities polling, code download, enter custom code, security
- Documentation better, more examples
- Testing
- Release to pypi

Work completed: 60 - 65 percent









Project Demo

DEMO









Problems encountered

- 1. pySpark program takes longer time than a normal Python program to run GA
- Test on very large datasets
- Improve the dag?
- Need to work on cluster?









Problems encountered

2. Population Size: experiment with same size and increasing size on Spark?









Any Questions?









Thank You