Generisanje random grafova

Anđela Damnjanović

Univerzitet u Beogradu, Matematički fakultet

7. januar 2024.



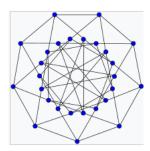
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Uvod

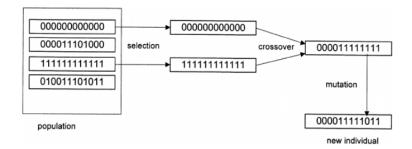
Uvod

- Značaj grafova praktični i teorijski
- Grafovski algoritmi
- Mane algoritama



Opis problema

- Željena svojstva → graf
- Genetski algoritam



Aviz projekat



Implementacija

Učitavanje test primera

```
{"num nodes": 6,
  "average degree": 3.33,
  "clustering coefficient": 0.7,
  "num of connected components": 1,
  "transitivity": 0.75}
```

- Organizacija ulaznih vrednosti
- Poziv algoritma
- Rezultati



Implementacija

Uvod

Klasa Individual

```
self.graph = nx.erdos_renyi_graph(self.num_nodes, 0.5)
```

```
return -((abs(num_of_graph_nodes - self.num_nodes))/max_nodes +
    abs(graph_avg_degree - self.avg_degree)/max_degree + abs(
    graph_clustering_coefficient - self.avg_clustering_coeff) +
    abs(graph_transitivity - self.transitivity)+ (abs(
    graph_connected_components - self.num_components)/
    max_components))
```

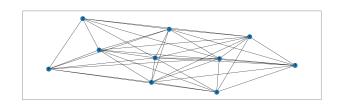
Implementacija

• Funkcije mutation i selection

```
1000 nx.to_numpy_array(parent1.graph)
   nx.from_numpy_array(adj_mat_cld1)
```

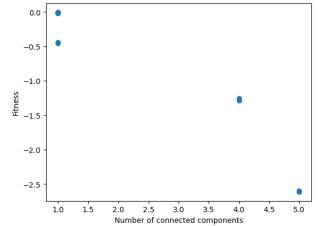
Tabela 1: Vrednosti fitness funkcije

Broj čvorova	Veličina populacije	Fitness
6	50	-0.005622222222222093
6	100	-0.005622222222222093
10	50	-0.02
10	100	-0.01599999999999997
10	50	-0.01085664682539699
10	100	-0.006736282964629461
25	50	-1.2832926628926629
25	100	-1.2528776855378152
50	50	-0.45898283539623946
50	100	-0.44200014007646093
75	50	-0.015890695284337637
75	100	-0.007531980904111335
100	50	-2.597159336271236
100	100	-2.6122425497135984





- Poboljšanja?
 - Postojanje linearnog modela?



Uvod

Poboljšanja?

• Matrica korelacije?

transitivity

clustering coeff

```
Correlation matrix is :
```

```
fitness
                             num of nodes
                                            ava dearee
                                                         num components
fitness
                   1.000000
                                 -0.658670
                                              0.323508
                                                              -0.958681
num of nodes
                  -0.658670
                                  1.000000
                                              0.442679
                                                               0.521044
avg degree
                   0.323508
                                  0.442679
                                              1.000000
                                                              -0.354475
num components
                  -0.958681
                                  0.521044
                                             -0.354475
                                                               1.000000
transitivity
                   0.615751
                                 -0.329264
                                              0.343455
                                                              -0.602434
clustering coeff
                  0.608040
                                 -0.311776
                                              0.362305
                                                              -0.594298
                   transitivity
                                 clustering coeff
fitness
                       0.615751
                                          0.608040
num of nodes
                      -0.329264
                                         -0.311776
avg degree
                       0.343455
                                          0.362305
num components
                      -0.602434
                                         -0.594298
```

1.000000

0.998739

0.998739

1.000000

- Poboljšanja?
 - Smanjenje verovatnoće za generisanje grana?

Tabela 2: Rezultati smanjenja verovatnoće za generisanje grana			
Broj čvorova	Veličina populacije	Fitness	
6	50	-0.005622222222222093	
6	100	-0.005622222222222093	
10	50	-0.1395151515151515	
10	100	-0.01200000000000000002	
10	50	-0.008549953314659122	
10	100	-0.005387076615423028	
25	50	-0.6687809523809524	
25	100	-0.48526616541353385	
50	50	-0.420012338376997	
50	100	-0.46940357205102456	
75	50	-0.0125517293241569	
75	100	-0.012004898528627055	
100	50	-1.428400037947298	
100	100	-2.6324747759596843	

- Poboljšanja?
 - Povećanje verovatnoće za mutaciju?

Broj čvorova	Veličina populacije	Fitness
6	50	-0.005622222222222093
6	100	-0.0611777777777778
10	50	-0.4827692307692308
10	100	-0.21071794871794872
10	50	-0.0360000000000000136
10	100	-0.006736282964629461
25	50	-1.4037338762346439
25	100	-1.283509261380821
50	50	-0.4757752270329916
50	100	-0.44939682755602156
75	50	-0.0099214931224301
75	100	-0.013052419602693637
100	50	-2.595627933058239
100	100	-2.6392986448062357

Zakljućak

- Relativno dobri rezultati
- Problem sa međusobno zavisnim svojstvima
- Prostor za poboljšanje

Literatura

Uvod

- [1] Annu Lambora, Kunal Gupta, and Kriti Chopra. Genetic algorithm- a literature review. 2019 International Conference on Machine Lear- ning, Big Data, Cloud and Parallel Computing (Com-IT-Con), 2019.
- [2] Vesna Marinkovic, Filip Maric, Strahinja Stanojevic, and Sana Stojanovic-Durdevic. Konstrukcija i analiza algoritama. Matematicki fakultet, Univerzitet u Beogradu, 2019.
- [3] M.E.J. Newman. Random graphs with clustering. Physical review letters, 2009.
- [4] Optimization algorithms. Complexica, 2023.
- [5] Vili Podgorelec, Janez Brest, and Peter Kokol. Power of heterogeneous computing as a vehicle for implementing e3 medical decision support systems.

