



















Table 4. Comparison of Prioritized GA-PSO with GA and Round-Robin

Parameter	with GA	with Round-Robin	Impact
Average CPU Utilization	+0.54%	+12.51%	Improvement
Execution Delay	-8.73%	-3.90%	Improvement
Average Waiting Time	-22.65%	-21.99%	Improvement
Energy Consumption	-17.81%	-1.68%	Improvement

From the result analysis, it is clear that the proposed algorithm allocates the resources to tasks in fog environment in efficient and effective manner, so as to minimize the delay, waiting time and energy consumption and maximize the CPU utilization. Hence, the proposed algorithm achieves the objective of allocating best and suitable resources to the client requests immediately in optimal and efficient way.

### 5. Conclusion

Fog computing is an emerging paradigm that offers storage and computation facility resources at the proximity of end devices. With the increasing prevalence of fog computing, the resource allocation to end user requests has become a relevant research issue. It aims for achieving the minimized resource wastage, execution delay, waiting time as well to save energy. Thus, a hybrid P-GA-PSO resource allocation algorithm has been proposed. The quantitative results of proposed algorithm were compared with Round-Robin and GA algorithms. The task allocation done using proposed algorithm showed reduced delay, waiting time and energy consumption by 8.73%, 22.65% and 17.81% respectively as well as improved resource utilization by 0.54% in comparison to GA. Similarly, the proposed algorithm when compared with Round Robin algorithm showed reduced delay, waiting time and energy consumption by 3.90%, 21.99% and 1.68% respectively as well as improving resource utilization by 12.51%. Further, a quantitative analysis presented in this work showed improved performance of proposed P-GA-PSO.

### References

- [1] Aazam M. and Huh Eui-Nam, (2015): Dynamic resource provisioning through Fog micro datacenter, IEEE International Conference on Pervasive Computing and Communication Workshops (PerCom Workshops), St. Louis, MO, pp. 105–110, doi: 10.1109/PERCOMW.2015.7134002.
- [2] Agarwal S., Yadav S., and Yadav A. K., (2015): An architecture for elastic resource allocation in Fog Computing, vol. 6, no. 2, pp. 7.
- [3] Bitam S., Zeadally S., and Mellouk A., (2018): Fog computing job scheduling optimization based on bees swarm, Enterp. Inf. Syst., vol. 12, no. 4, pp. 373–397, doi: 10.1080/17517575.2017.1304579.
- [4] Bittencourt L. F., et. al. (2017): Mobility-Aware Application Scheduling in Fog Computing, IEEE Cloud Comput., vol. 4, no. 2, pp. 26–35, doi: 10.1109/MCC.2017.27.
- [5] Deb K., et. al., (2002): A fast and elitist multiobjective genetic algorithm: NSGA-II, IEEE Trans. Evol. Comput., vol. 6, no. 2, pp. 182–197, doi: 10.1109/4235.996017.
- [6] Deng R., et. al., (2016): Optimal Workload Allocation in Fog-Cloud Computing Towards Balanced Delay and Power Consumption, IEEE Internet Things J., vol. 3, no. 6, pp. 1171–1181, doi: 10.1109/JIOT.2016.2565516.
- [7] Ghobaei-Arani M., Sourii A., and Rahmian A. A., (2019): Resource Management Approaches in Fog Computing: A Comprehensive Review, J. Grid Computing, vol. 18., pp.1-42 doi: 10.1007/s10723-019-09491-1.
- [8] Hu P., et.al., (2017): Survey on fog computing: architecture, key technologies, applications and open issues, J. Netw. Comput. Appl., vol. 98, pp. 27–42, doi: 10.1016/j.jnca.2017.09.002.
- [9] Kennedy J. and Eberhart R., (1995): Particle swarm optimization, Proceedings of ICNN'95 - International Conference on Neural Networks, Perth, WA, Australia, vol.4, pp. 1942-1948, doi: 10.1109/ICNN.1995.488968.
- [10] Khurram, M., Biradar Dr., A., (2020): A link based genetic algorithm approach in optimizing routing in wireless adhoc network, Indian J. Comput. Sci. Eng. 11, pp. 488–496. <https://doi.org/10.21817/indjce/2020/v11i5/201105112>
- [11] Kochar V. and Sarkar A., (2016): Real time resource allocation on a dynamic two-level symbiotic fog architecture, Sixth International Symposium on Embedded Computing and System Design (ISED), Patna, India, pp. 49–55, doi: 10.1109/ISED.2016.7977053.
- [12] Konak A., Coit D. W., and Smith A. E., (2006): Multi-objective optimization using genetic algorithms: A tutorial, Reliab. Eng. Syst. Saf., vol. 91, no. 9, pp. 992–1007, doi: 10.1016/j.res.2005.11.018.
- [13] Kumar D. and Raza Z., (2015), A PSO Based VM Resource Scheduling Model for Cloud Computing, IEEE International Conference on Computational Intelligence & Communication Technology, Ghaziabad, India, pp. 213–219, doi: 10.1109/CICT.2015.35.
- [14] Kumar M. and Sharma S. C., (2019): PSO-based novel resource scheduling technique to improve QoS parameters in cloud computing, Neural Comput. Appl., vol.32, pp. 12103-12126, doi: 10.1007/s00521-019-04266-x.
- [15] Mahmud R., Kotagiri R., and Buyya R., (2018): Fog Computing: A Taxonomy, Survey and Future Directions, in Internet of Everything, B. Di Martino, K.-C. Li, L. T. Yang, and A. Esposito, Eds. Singapore: Springer Singapore, pp. 103–130.
- [16] Mohan N. R. R. and Raj E. B., (2012): Resource Allocation Techniques in Cloud Computing -- Research Challenges for Applications , Fourth International Conference on Computational Intelligence and Communication Networks, Mathura, Uttar Pradesh, India, pp. 556–560, doi: 10.1109/CICN.2012.177.
- [17] Name H. A. M., Oladipo F. O., and Ariwa E., (2017): User mobility and resource scheduling and management in fog computing to support IoT devices, Seventh International Conference on Innovative Computing Technology (INTECH), Luton, pp. 191–196, doi: 10.1109/INTECH.2017.8102447.
- [18] Ni L., et. al, (2017): Resource Allocation Strategy in Fog Computing Based on Priced Timed Petri Nets, IEEE Internet Things J., vol. 4, no. 5, pp. 1216–1228, doi: 10.1109/JIOT.2017.2709814.
- [19] Rahbari D. and Nickray M., (2017): Scheduling of fog networks with optimized knapsack by symbiotic organisms search, 21st Conference of Open Innovations Association (FRUCT), Helsinki, pp. 278–283, doi: 10.23919/FRUCT.2017.8250193.
- [20] Sun Y. and Zhang N., (2017): A resource-sharing model based on a repeated game in fog computing, Saudi J. Biol. Sci., vol. 24, no. 3, pp. 687–694, doi: 10.1016/j.sjbs.2017.01.043.