





## 5. Methodology

In this section, the introduction of Fog computing within the context of an IoT network is demonstrated through a PoC using the network simulation software toolset to demonstrate Fog computing advantages and how these advantages address Cloud computing limitations such as high energy consumption, latency and network bandwidth usage.

For the benefit of this paper, the network simulation software toolset evaluation exercise is conducted to select the best simulator to use as part of the method to model Fog and Cloud computing within an IoT network with a focus on energy consumption challenge and how Fog computing addresses this challenge.

### 5.1 Network Simulation Software Toolset

Various network simulation software toolsets with a number of features, functionalities, characteristics and input topology formats have been developed in line with the evolution of IoT, Cloud, Edge and Fog computing paradigms [14]. However, each toolset has advantages, disadvantages and trade-offs involved, which limits the capabilities of one toolset satisfying requirements for every type of network infrastructure scenario.

Testing in a real-world large scale network infrastructure is not practical, high-cost, with risk to impacting live services and applications. Therefore, using a network simulator software toolset to simulate a testbed or development environment is necessary and cost effective within large-scale network infrastructures to test novel solutions prior to launching in a production or live environment [15]. Consequently, determining the most suitable and appropriate network simulation tool can be quite a daunting task for researchers and could lead to severe delays, limitations or flawed simulations.

The simulator used to model and evaluate Cloud and Fog computing continuums should have certain characteristics and capabilities such as measuring latency, energy consumption and network bandwidth usage to fully represent realistic environments reflective of the outputs within a complex environment.

Completing the first stage of the evaluation exercise, three network simulation software toolsets were identified and chosen NetFogSim++, iFogSim and CupCarbon. Some of the attributes validated ranged from extent of programming to energy consumption measurement model inclusion as a function.

The Table 1 presents three network simulation software toolsets evaluated from literature which meet the requirement for energy consumption module attributes which is critical to evaluating infrastructure data management and energy efficiency.

Table 1. Network simulation software toolset evaluation table

Attributes	FogNetSim++	iFogSim	CupCarbon
Computing Paradigm (Target System)	Fog computing	Fog computing	Edge computing
Infrastructure and Network Level Modelling	Distributed data centres, sensors, Fog nodes, Broker, Network links, Delays, Handover, Bandwidth.	Cloud data centres Sensors Actuators Fog Devices Network links Delay Network usage Energy consumption	Sensors, Fog nodes, Network links, Bandwidth, Network Protocols
Application Level Modelling	Fog network	Data stream, Stream processing	IoT, WSN
Mobility	Yes	Yes	Yes
Scalability	Yes	Yes	Yes
Graphical User Interface	Yes	Yes	Yes
Energy Consumption Module	Energy consumption limited to Edge node energy measurement	Energy consumption measured across all IoT network layers by different classes of devices, sensors, and actuators	Energy consumption measured as a function of simulated time
Sensor Simulation	Yes	Yes	Yes, however, not on all protocol layers
Physical Sensors	Yes	Yes	Yes
Data Management Process	Yes	Yes	Yes
Capabilities			
Network Communication	Yes	Yes	Yes
Mobile Nodes	Yes	Yes	Yes

## 6. Fog Computing Network Simulator Software Benefits and Challenges

The following are some of the benefits of Fog computing network simulation software toolsets; the support of multiple features such as fault injection and the arbitrary removal of links and nodes, this allows for the variable mechanism tests for resilience, fault tolerance and complex experiments within realistic environments [15]. Most simulators also have monetary and energy consumption possibilities embedded within network simulator software toolsets. While other network simulators especially for Fog computing has dynamic capabilities such as variable service infrastructure, inclusion and exclusion of more resources on demand to support the dynamic nature of Fog computing environments [15]. There are more benefits to using a network simulator software toolset however, as a primary focus of this paper, most of the benefits mentioned are in line with and supports Fog computing environments.

The amplified mobile device ubiquity and the surge of network and computing end points over the decade through low cost sensors connections has certainly been on the increase. The traditional Cloud computing network infrastructure was not designed with the capacity to manage compute, processing and storage for billions of devices with multiple end points and irregular network connection in a dynamic environment [16]. In addition, the profound complexity of design, test, implementation and heterogeneous formats of data sources from the sharp increase in number of sensors embedded within IoT devices will require a greater level of understanding and a steep learning curve to be able to model and simulate these environments. These challenges through lessons learnt serves to encourage the developmental review of a more improved design, reliability, fault tolerance and holistic network simulation software toolset modelling and framework.

Additionally, current network simulation software toolsets have to manage a multitude of variable workloads due to increased number of devices consequently, there should be greater support for modelling heterogeneous data abstractions and network topological workflow, heterogeneous query and dataflows.

FogNetSim++ is a network simulation software toolset designed on the open source OMNeT++ framework and developed for Fog computing, enabling users to configure large scale Fog network simulations using detailed configuration options, extensive libraries and discrete events. Furthermore, FogNetSim++ allows users to simulate variable features of heterogeneous devices, such as handover to track data or device source using flexible design and the incorporation of bespoke algorithms through class extension [17]. Some key features of the network simulation software toolset evaluation specific include the energy consumption module which FogNetSim++ supports through extensive energy measure for each participating Fog nodes and mobile devices, and a range of mobility models.

iFogSim is a Java technology based state-of-the-art simulation software toolset designed for Fog computing environment scenario experiments in assessing resource management, scheduling policies including impact on energy consumption, latency and network bandwidth usage measurement [18]. Furthermore, iFogSim has extensive capabilities to simulate and model various infrastructure scenarios and features such as RAM, transient storage capabilities, CPU and energy consumption modelling. As a Java based network simulation software toolset, iFogSim supports multiple java classes that corresponds to applications, sensors, actuators, tuples and Fog devices, Fog nodes and large-scale heterogeneous nodes. In addition, iFogSim has been extended to optimise data placements in IoT devices and Fog nodes to ensure the management of data placements in relation to various objectives such as improving energy consumption, latency and network bandwidth usage [19]. iFogSim, was designed with detailed energy consumption measurement capabilities across all layers of the network which is the focus of the PoC. Consequently, iFogSim is the best fit for measuring energy efficiency, latency and network bandwidth usage in an IoT network infrastructure.

CupCarbon is a Java based developed network simulation software toolset designed with easy to use configurable objects, multi agent capabilities with embedded wireless network simulator geolocation sensor which supports digital geographic interface such as OpenStreetMap and simulate sensor networks and modelling. CupCarbon main architecture comprises of about seven modules: 2D/3D city model module, Mobility module, Network module, Communication Script module, Radio Channel

Propagation module, Interference module and Simulation module [20]. The mobility, network, communication and simulation modules are relevant however, there are significant limitations such as the energy simulation module which is calculated as a function of simulated time and not an extensive analysis of the data management process. Furthermore, CupCarbon is developed and designed for Edge computing and does not support the Fog computing layer specific technology paradigm. As a result, CupCarbon is not a suitable network simulation software toolset for this paper.

Under NetFogSim++, the energy measurement is done through participating Fog nodes only excluding the sensor and actuator embedded device layer which limits the scale at which this simulator can model Fog capabilities while in iFogSim, energy consumption is measured across all the devices including Fog nodes, applications, tuples and sensors across all network computing layers. Where iFogSim provides a comprehensive and extensive energy consumption analysis through all computing layers involved within the infrastructure. However, with CupCarbon, energy consumption is measured as a function of the time it takes to transmit data to the Cloud computing layer. This lacks support for sensors, actuators, geographical mobility and other Fog modelling capabilities. Therefore, iFogSim network simulation software toolset is best suited to evaluate data management and energy consumption for Cloud and Fog computing infrastructures.

## 7. Energy Consumption Estimation Analysis

The simulation scenario modelled in Figures 2 and 3 is configured with data processed at 3 million instructions per second, uplink bandwidth of 1GB, downlink bandwidth of 1.5GB while the sensors and actuators transmit data to the Fog node at the rate of 50 milli seconds.

iFogSim has a network GUI which visualises the topological models as shown in Figure 2 and 3 with the required parameters used to demonstrate the PoC. Furthermore, Figure 4 highlights energy consumption increase for Cloud data centre when data processing is routed to the Cloud while the opposite occurs when processing is routed through the Fog nodes (DC). However, the rate of change for DC Cloud and DC Fog also varies, with the level energy consumption increased for both DC Fog and DC Cloud as the configuration is scaled up however, at different rates as shown in Figure 4. The network topological configuration of 5 Fog nodes and 3 devices (5,3) with an energy output measurement of 8072KJ and 16145KJ for DC Fog and DC Cloud respectively. This shows the rate of energy consumption increase for DC Cloud is roughly double the rate of increase for DC Fog. With regards to scalability, Figures 2 and 3

describe increased scale of Fog nodes and devices from 1,3 to 5,5 and proportionate to an increase in energy consumption.

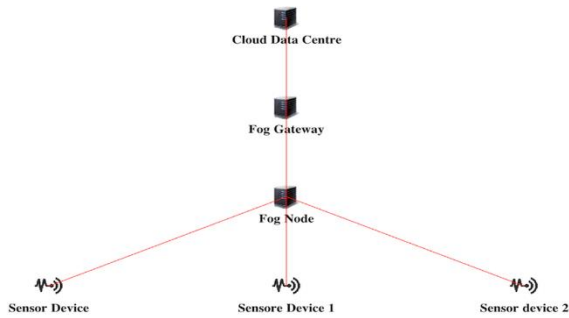


Figure 2. iFogSim network topology configuration 1 Fog node with 3 devices connected (1,3)

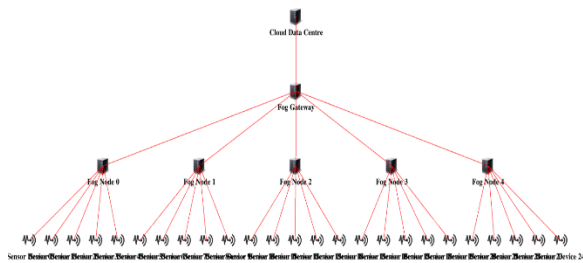


Figure 3. iFogSim network topology configuration 5 Fog nodes with 5 devices connected (5,5)

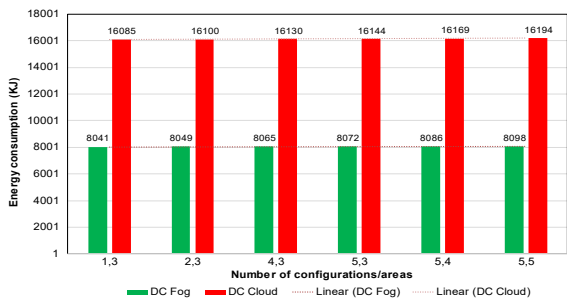


Figure 4. DC Fog and Cloud energy consumption combined output

## 8. Results and Discussion

The results demonstrate that Fog infrastructure is significantly more energy efficient than Cloud approaches when processing data and as network infrastructure scales. Based on linear trend analysis in Figures 5 and 6, it can be observed that the energy consumption linear equations are  $Y=11.589x+8027.8$  and  $Y=21.916x+16061$  for DC Fog and DC Cloud.

The linear equations gradient parameter indicates the energy consumption rate of change as infrastructure increases. DC Fog and DC Cloud demonstrated energy consumption rate of change values of 11.589 and 21.916 respectively. Although

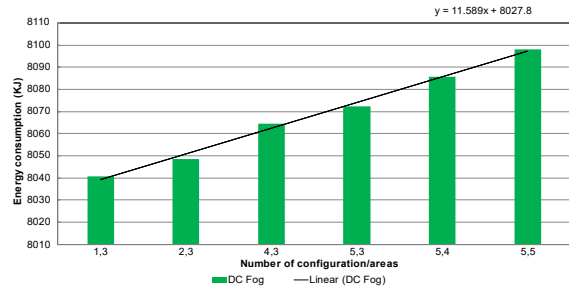


Figure 5. DC Fog energy consumption output

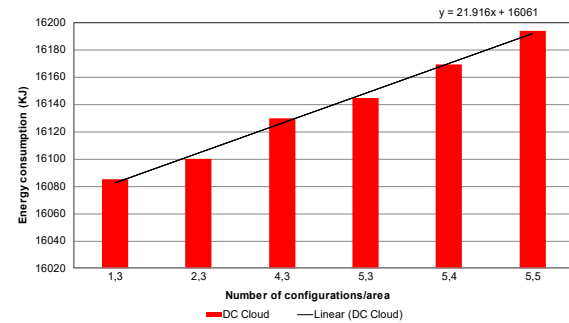


Figure 6. DC Cloud data centre energy consumption output

performance parameters demonstrate that both DC Fog and DC Cloud are linear. The rate of change for energy consumption in DC Fog data processing is 89% lower compared to DC Cloud. Furthermore, as scale increases, DC Cloud increases at a higher rate of change of 62%, making DC Fog a preferred option for energy savings. However, due to limited capabilities, only 5 user devices with up to  $5 \times 10000 = 50000$  concurrent tasks can be computed at a single Fog node compared to DC Cloud based approach which is more scalable and can handle up to 30 end user devices with  $30 \times 10000 = 300000$  tasks per node due to increased range of capacity.

## 9. Conclusion

The network simulation software toolset evaluation demonstrated iFogSim as the most suitable simulator toolset for investigating the energy consumption measurements for data management in an IoT network infrastructure. The PoC further demonstrates that the introduction of Fog computing layer to the IoT architecture extending the traditional Cloud computing infrastructure successfully reduces the energy consumption levels when data is processed through the Fog computing layer as opposed to the Cloud computing layer. There are other benefits of Fog computing alongside the various methods that can further reduce energy consumption and conform to some Sustainable Development Goals (SDG) such as improved scheduling in Fog nodes and the use of algorithms to enhance data management performance.

The results of this paper can be expanded for further investigations into critical areas that can further improve energy efficiency with Fog computing and algorithms which can further improve latency and network bandwidth usage. Further studies should be conducted in these areas to reduce the impact of the plethora of smart IoT devices as they continue to grow at an exponential rate.

## 10. References

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