

# CSE 511: Data Processing at Scale

Amey Bhilegaonkar  
Arizona State University  
abhilega@asu.edu

## Introduction

In this project, Kubernetes, Kafka, Docker, and Neo4j are all combined to provide a highly scalable and available data processing pipeline. The pipeline will receive streaming data produced by Kafka stream, process it, and transfer it to Neo4j for analytics and processing in close to real-time. This project involved stepwise execution wherein the first step involved setting up the Kafka and Apache Zookeeper together using the orchestrator minikube. Data from the parquet file will be ingested and distributed to other services via Kafka. In the second step, integration of Neo4j into the setup is performed, and the data is streamed into Neo4j for further analysis and processing. Once the data is ingested into Neo4j, the analysis and algorithms such as PageRank and BFS were applied to graph-based data to explore the data.

## Methodology

The aim of this section is to propose a methodology for implementing Kafka and Neo4j as a service using Helm and Minikube in Kubernetes. The proposed methodology consists of the following steps:

### 1.1. Environment Setup

The first step in installing Kafka and Neo4j as a service is to create the proper environment. The installation of Minikube, Kubernetes, and Helm is the first step towards putting the Kafka and Neo4j services into practice. A local Kubernetes cluster called Minikube offers the essential conditions for setting up and running the services. Helm, on the other hand, is a Kubernetes package manager that streamlines the cluster's application administration and installation processes.

**1.2 Kafka and Zookeeper Installation:** This can be achieved by creating a Kafka cluster and Zookeeper on Kubernetes. The two YAML files `kafka-setup.yaml` and `zookeeper-setup.yaml` contain the necessary configurations required for a basic Kafka cluster to get up and running.

**1.3 Neo4j Installation:** This step involves storing the data in Neo4j. For this purpose, an official Neo4j Helm chart can be utilized to create a Neo4j cluster. The `neo4j-values.yaml` and `neo4j-service.yaml` will help us get the Neo4j service up and running to serve as a backend database.

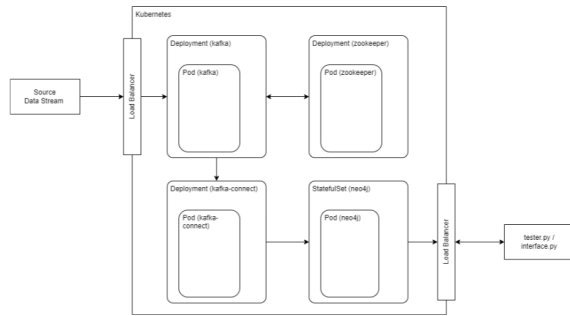
**1.4 Data Ingestion:** After the Kafka cluster is up and running, the next step is to ingest data into the Kafka cluster. For this step, we will run a `data_producer.py` file which will load the graph data from the parquet file to Kafka and then to Neo4j.

**1.5 Testing and Deployment:** In the end, thorough testing is required to guarantee the implementation's appropriate functionality. Simple tests like sending a message via Kafka-stream and receiving it on the other end should be included in the testing process. Also, when the data is loaded with `data_producer.py` into a graph database, we can run `tester.py` along with `interface.py` from phase 1 of this project to see if BFS and PageRank algorithms are working fine. When testing is successfully finished, the implementation can be reliably put into use.

**1.6 Testing on limited resources:** For testing purposes, since I had limited Memory and Cores for my personal laptop, it was really hard for me to run the original parquet file, it took almost a day for me just to run that file and still it was running the next day. So I decided to divide the whole file into chunks of 50 records and then test my code with 3 chunked files of 150 records. I modified the `tester.py` file, added my own test cases, and tested them. The part-parquet files and `tester.py` file is attached with the code. Also, screenshots from the results are attached to this report and the rest screenshots are pushed on [GitHub](#). Also, the Python file with which I created chunk files was also pushed to GitHub.

With the steps followed as described above, and with automation test cases created on my own, I learned new things and explored various resources to debug the issue which gave me exposure to various other extensions that are used to manage minikube and help it automate it.

## Results



## Architecture of services

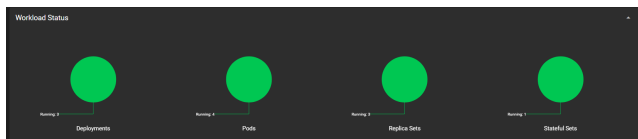
After the successful deployment of pods of Zookeeper, Kafka, Kafka-connector, and Neo4j we should see the following results with the commands to test our flow from point A to point B.

The following Commands were run to check if our deployments are correct:

1. `kubectl apply -f ./zookeeper-setup.yaml`
2. `kubectl apply -f ./kafka-setup.yaml`
3. `helm install my-neo4j-release neo4j/neo4j -f neo4j-values.yaml`
4. `kubectl apply -f neo4j-service.yaml`
5. `kubectl apply -f Kafka-neo4j-connector.yaml`
6. `kubectl port-forward svc/neo4j-service 7474:7474 7687:7687`
7. `kubectl port-forward svc/Kafka-service 9092:9092`

### For Testing:

1. `kubectl get pods`
2. `kubectl get services`
3. `python3 data_producer.py`
4. `python3 tester.py`



## Kubernetes Deployments

Services						
Name	Labels	Type	Cluster IP	Internal Endpoints	External Endpoints	Created
mysql-service		ClusterIP	10.102.84.196	mysql-service:3306 TCP mysql-service:3306 UDP mysql-service:3306 TCP		25.03.2018 ago
mysql-volume	app=mysql-database app.kubernetes.io/managed-by=k8s mysql-volume/mysql-volume:my-mysql-volume	ClusterIP	10.102.28.146	mysql-volume-mysql:3306 TCP mysql-volume-mysql:3306 UDP mysql-volume-mysql:3306 TCP mysql-volume-mysql:3306 TCP		26.03.2018 ago
	mysql-volume					
mysql-volume-admin	app=mysql-database app.kubernetes.io/managed-by=k8s mysql-volume/mysql-volume:my-mysql-volume	ClusterIP	10.102.108.21	mysql-volume-admin:3306 TCP mysql-volume-admin:3306 UDP mysql-volume-admin:3306 TCP mysql-volume-admin:3306 TCP		26.03.2018 ago
	mysql-volume					
mysql-database-mysql	app=mysql-database app.kubernetes.io/managed-by=k8s mysql-volume/mysql-volume:my-mysql-volume	LoadBalancer	10.10.272.206	mysql-database-mysql:3306 TCP mysql-database-mysql:3306 UDP mysql-database-mysql:3306 TCP mysql-database-mysql:3306 TCP mysql-database-mysql:3306 TCP		26.03.2018 ago
	mysql-database					
kafka-service		ClusterIP	10.111.33.84	kafka-service:9092 TCP kafka-service:9092 TCP kafka-service:9092 TCP		27.03.2018 ago
kafka-service	app=kafkaconnect-service connectors=kafkaconnect	Headless	10.137.4.131	kafkaconnect-service:2181 TCP kafkaconnect-service:8161 TCP		28.03.2018 ago
kafka-service	app=kafkaconnect-service connector=kafkaconnect	ClusterIP	10.146.1.1	kafkaconnect-connector:3306 TCP kafkaconnect-connector:3306 TCP		01.04.2018 ago

## Kubernetes Services

Replica Sets					
Name	Images	Labels	Pods	Created	T
kylo-converter-1000000	vanke/kilo-convert	app=kylo-convert pod-template-hash=1000000	1/1	22/00:00	
kylo-converter-1000000	confounding/kylo-7.3	app=kylo pod-template-hash=1000000	1/1	20/00:00	
kylo-converter-1000000	confounding/convertor-7.3	app=convertor pod-template-hash=1000000	1/1	20/00:00	

Starful Sets					
Name	Images	Labels	Pods	Created	T
kylo-convert-1000000	vanke/kilo-convert	app=kylo-convert	1/1	20/00:00	
kylo-converter-1000000	confounding/kylo-7.3	app=kylo-convert pod-template-hash=1000000	1/1	20/00:00	
kylo-converter-1000000	confounding/convertor-7.3	app=convertor pod-template-hash=1000000	1/1	20/00:00	

## Kubernetes Replica Status

Deployments									
Name	Images	Labels	Pods	Created					
helloworld	vestika/hello-world-connect	app: helloworld	1/1	21 minutes ago					
helloworld-deployment	cofounders/hello-world	app: helloworld	1/1	21 minutes ago					
helloworld-deployment	cofounders/hello-world	app: helloworld	1/1	28 minutes ago					

Pods									
Name	Images	Labels	Node	Status	Restart	QoS (Usage Limit)	PRIORITY (Type)	Created	
helloworld-5454545454	vestika/hello-world-connect	app: helloworld pod-template-hash: 5454545454	minikube	Running	0			21 minutes ago	
helloworld-5454545454	vestika/hello-world-connect	app: helloworld pod-template-hash: 5454545454	minikube	Running	0			24 minutes ago	
helloworld-5454545454	vestika/hello-world-connect	app: helloworld pod-template-hash: 5454545454	minikube	Running	0			28 minutes ago	

## Kubernetes Pod and Deployment Status

The running pods show that our services are up and running. Now once we login into <http://localhost:7474/browser/> we can see that our graph database Neo4j is up. We can run the following command to see if the database got updated with the right data or not:

1. `match(n) return(n)` - This will give us all the nodes present in the graph



### Graph Visualization of one of the Chunk files

Since I had limited hardware resources, I had to divide the big parquet file into small chunks and then test it. I ran the following commands to run `custom_data_producer.py` to load the data from the parquet file into the database and then ran `custom_tester.py` to verify the results.

1. `cd CustomTests\`
2. `python custom_data_producer.py`
3. `python custom_tester.py`

The below screenshot shows the output of the above command.

```
C:\Amev\ASU\DPSAssignments\project-2-Phase3\abhilega-project-2\CustomTests>python custom_tester.py
Trying to connect to server
Server is running

-----
Testing if data is loaded into the database
Count of Edges is correct: PASS
Count of Edges is correct: PASS

-----
Testcase: 1 for PageRank
PageRank Test 1: PASS
Deleting graph!

-----
Testcase: 2 for PageRank
PageRank Test 2: PASS
Deleting graph!

-----
Testcase: 3 for PageRank
PageRank Test 3: PASS

-----
Testing if BFS is working
BFS Test 3: PASS

-----
Testing if BFS is working
BFS Test 3: PASS

-----
Testing if BFS is working
BFS Test 3: PASS

-----
Deleting graph!
Deleting Nodes and Relations!
```

#### The output of the Test Cases

This command will tell us if test cases for algorithms BFS and PageRank ran fine. Everything was successful.

For this particular project, I created small subsets of the original parquet file and tested it out with that file creating my own test cases, the corresponding tester.py file is submitted along with the other codes [here](#).

## Contributions

In this section, I will be talking about my contributions to the project that I completed as part of this course. The following section will describe my contributions and learnings from the project.

### Individual Project - [Project 2](#):

This project was to test and expand our understanding of Streaming messaging services like Kafka, a graph database management system like Neo4J, and DevOps technologies such as Docker, and Kubernetes.

This project was divided into two phases - phase-1 and phase-2. My key contribution to the project was:

**Phase 1:** Implemented a containerized application to load the dataset. This was done by running a docker container which was responsible for installing necessary dependencies and downloading the parquet file converting it to a CSV file and then I wrote the Neo4J queries to load the data from the file to the database. Throughout these queries, a certain relation amongst the node was added along with the addition of the nodes.

Once the [data loading](#) was done, I installed GDS (graph data science) plugin for Neo4J in the docker container. Once I had all the plugins and dependencies installed, I

added two key components of this project, which were graph algorithms to explore the data:

1. [Breadth First Search](#)
2. [PageRank](#)

These algorithms helped explore the graphs data and the corresponding code can be found [here](#).

### [Phase 2](#):

For this phase, I implemented containerized apps for various services on minikube / kubernetes. I successfully communicated between the services and loaded data from point A to point B, flowing through all of the services. I implemented YAML files on minikube to get the following services up and running:

1. Zookeeper Service
2. Kafka-setup Service
3. Neo4j Service
4. Kafka-connector Service

Apart from the services and asked task, I implemented an automation grading script for my own since one was not provided with the assignment and it can be found [here](#). I also created well-written [documentation](#) instructing steps that needed to be followed to run the project and perform custom test cases.

## Discussion and Learnings

This project involved a lot of learning for me and the curve was definitely exponential. I got to learn about Kubernetes, and how the applications are scaled and managed by the orchestration tools such as Kubernetes. Kubernetes plays an important role in scaling applications, and it is crucial for managing containers in a production environment. Through this project, I got to learn about the different Kubernetes objects such as pods, deployments, and services, and how they all work together to ensure that the application is running smoothly.

I also got to learn how streaming applications work together with services like Kafka, which is used to build real-time data pipelines and streaming applications. In simple terms, it is a tool that allows you to send and receive large volumes of data between different systems or applications in real time. Kafka is a distributed streaming platform created to manage large amounts of data quickly and efficiently, making it an essential tool in a variety of sectors like banking, healthcare, and retail.

I liked one thing about this project: everything was pointed in a direction in which we could step and then gather information on our own, and apply that knowledge here to build scalable containerized applications, nothing was spoon-fed. Because of this, I got to explore technologies that I was not aware of, starting from Docker, Neo4j,

Kafka, Kubernetes, Minikube, etc. Through this project, I was able to improve my problem-solving abilities and gain more self-confidence when experimenting with new technology.

I think, for me, future work on this project will be significantly important, to take this project to AWS, a concept introduced in the class, and I would really like to explore that option as well. I will get to learn and explore AWS services, and how everything works in integration with one another in a cloud-based environment.

In conclusion, this project gave me insights into different technologies and services used for streaming jobs in real life, with a high learning curve. It helped me to develop a better understanding of how the different components of an application work together in a production environment. I also learned about the importance of message/streaming services such as Kafka, and how they can be used to build real-time data pipelines and streaming applications. Overall, this project was an excellent learning experience for me, and I am excited to continue exploring new technologies and services in the future.

## Conclusion

Through this project, I gained a deeper understanding of how data flows in real-time from source systems to databases, and the crucial role that message/streaming services play in facilitating this flow. I learned about Kafka and SQS, both of which enable the reliable and scalable transmission of data between applications and services. This knowledge will be invaluable in any future work I do with data-intensive applications.

Furthermore, I was able to delve into Neo4j, a graph database management system, and understand how it differs from traditional relational databases. I explored its unique features and capabilities and gained a greater appreciation for its usefulness in scenarios where relationships between data points are crucial. Overall, this project has given me a practical understanding of how complex applications are deployed to servers, and how containerized deployments enable multiple services to communicate and work together seamlessly.

## References

1. How to do WSL config to Manage the Docker Memory and CPU cores: [Microsoft Official Documentation](#)
2. [Official Kubernetes Documentation](#)
3. [Official Kubernetes Handbook](#)
4. YouTube [playlist](#) DevOps BootCamp to understand the depth of Docker and Kubernetes

5. Neo4J [Manual](#) for Kubernetes
6. Neo4J [documentation](#) for Kafka
7. Canvas [Discussion](#) section for this project