

Hybrid-Cloud Streaming Workshop

Table of Contents

Introduction

Lab 1: Connecting to your Workshop Environment

Lab 2: Getting Started

Lab 3: Stream CDC events to your Local Kafka Cluster

Lab 4: Stream events to Confluent Cloud

Lab 5: Building a Realtime Supply & Demand Application using KSQL

Lab 6: Getting Started with KSQL

Lab 7: Querying Streams using KSQL

Lab 8: Creating KSQL tables

Lab 9: KSQL Stream-to-Stream Joins

Lab 10: KSQL Stream-to-Table Joins

Lab 11: Streaming Stock Levels

Lab 12: Pull Queries

Lab 13: Streaming Recent Product Demand

Lab 14: Streaming "Out of Stock" Events

Lab 15: Replicate Events to On-Premise Kafka

Lab 16: Sink Events into MySQL

Optional Lab: Stream Sales & Purchases to Google Cloud Storage Optional Lab: Stream Sales & Purchases to Google Big Query Optional Lab: Stream Sales & Purchases to MongoDB Atlas

Wrapping up

Introduction

The popularity of Hybrid and Multi cloud architectures are on the rise as organizations continue to take advantage of cloud computing.

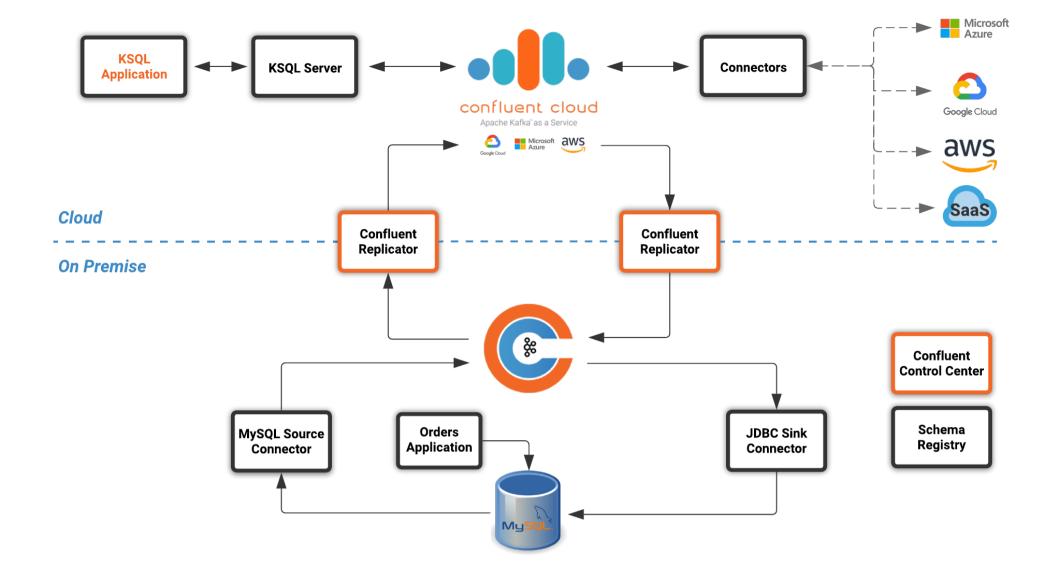
Some of the requirements driving these new modern architectures are as follows.

Organizations need to...

- Synchronize data between on-premise and the cloud
- Migrate data from on-premise into the cloud
- Synchronize data across multiple cloud providers to reduce risk
- Synchronize data across multiple cloud providers to avoid vendor lock-in
- Access the best-in-breed services across multiple cloud providers

In addition to Hybrid and Multi cloud architectures, organizations are also looking to become more event driven. The Confluent Platform is a streaming platform that can stream data, in real time, to the systems that need it, when they need it, across an entire organization. Processes that were once batch can now become real time, every event can be used to trigger other services and this can all be done using a common API with low latency and high throughput.

In this workshop we will explore how the Confluent Platform and Confluent Cloud can enable these architectures by building a real time supply and demand appplication using KSQL.



Lab 1: Connecting to your Workshop Environment

Your environment represents an on-premise data center and consists of a virtual machine hosted in the cloud running several docker containers. In a real world implementation, some of the components would be deployed differently but the logical data flow that we will be working on would remain the same.

To login to your virtual data center open a terminal session and use the credentials that were assigned to you

```
ssh dc01@34.89.105.9
```

Once logged in run the following command to confirm that you have several docker containers running

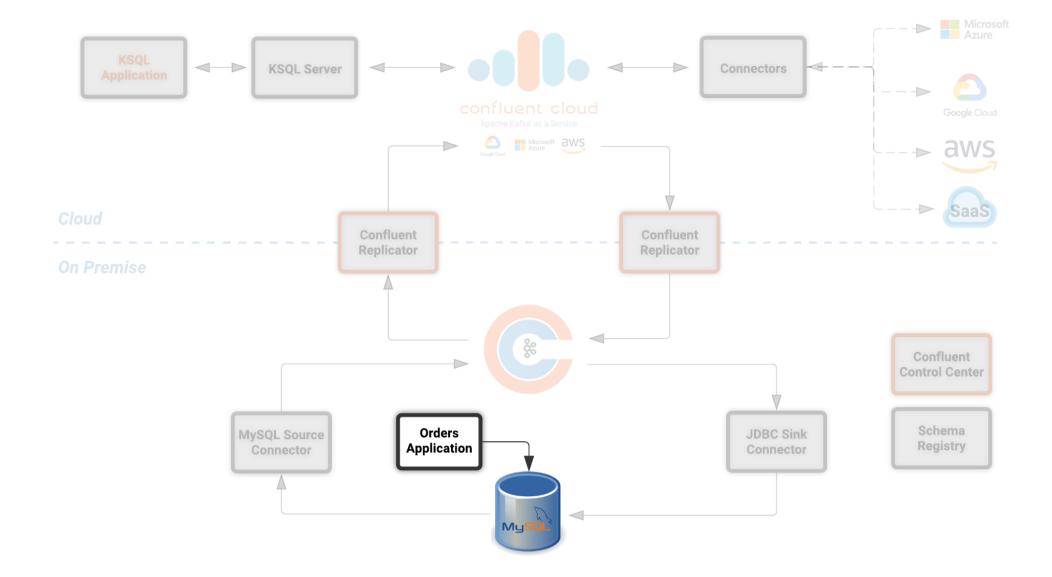
```
docker ps --format "table {{.ID}}\t{{.Names}}\t{{.RunningFor}}\t{{.Status}}"
```

You should see something similar to this:-

```
CONTAINER ID
                    NAMES
                                               CREATED
                                                                    STATUS
                                               6 minutes ago
d71e05e9b3b5
                    db-trans-simulator
                                                                    Up 6 minutes
9100108345db
                    mysql
                                               6 minutes ago
                                                                    Up 6 minutes (healthy)
                    ksql-cli
                                                                    Up 7 minutes
e280fd878cce
                                               7 minutes ago
                                                                    Up 7 minutes (healthy)
edcd99707a7a
                    ksql-server-ccloud
                                               7 minutes ago
d6ca56beb72e
                    control-center
                                               7 minutes ago
                                                                    Up 7 minutes
                    kafka-connect-ccloud
                                                                    Up 7 minutes
31bf790e76e7
                                               7 minutes ago
8b35d343e6d6
                    kafka-connect-onprem
                                               7 minutes ago
                                                                    Up 7 minutes
4aa6a7cd76c3
                    schema-registry
                                               7 minutes ago
                                                                    Up 7 minutes
                                               7 minutes ago
84022bbf75e5
                    broker
                                                                    Up 7 minutes
b0f2aefb2042
                    zookeeper
                                               7 minutes ago
                                                                    Up 7 minutes
                                               7 minutes ago
                                                                    Up 7 minutes
ee2983ac4bcc
                    workshop-docs-webserver
```

Lab 2: Getting Started

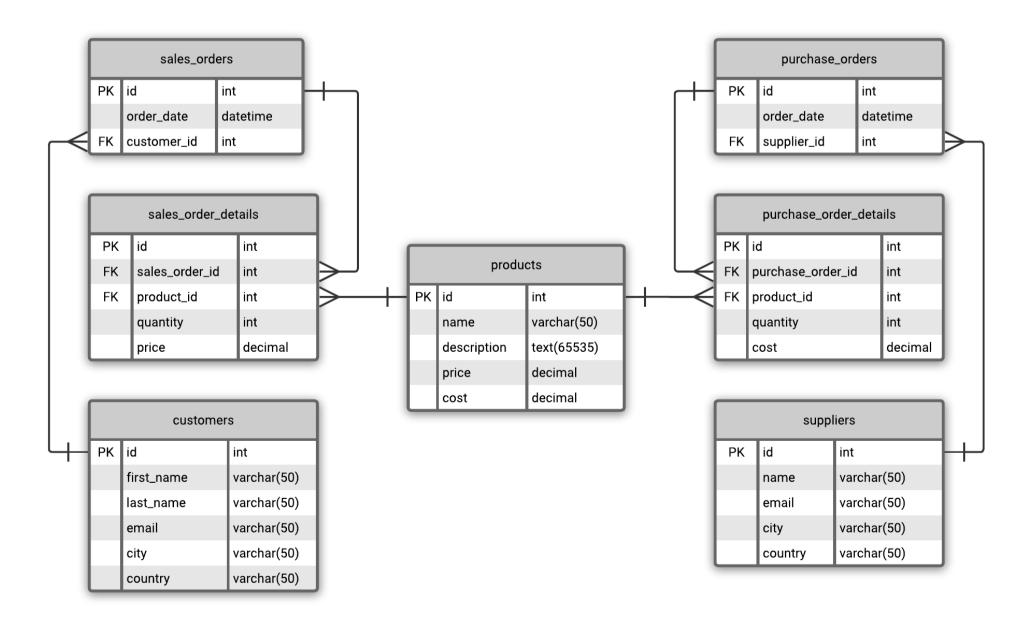
The data source for this workshop will be a MySQL database running in your data center. Connected to this database is a orders application which we will discuss shortly.



Database Schema

The MySQL database contains a simple schema that includes *Customer*, *Supplier*, *Product*, *Sales Order* and *Purchase Order* information.

The idea behind this schema is simple, customers order products from a company and sales orders get created, the company then sends purchase orders to their suppliers so that product demand can be met by maintaining sensible stock levels.



We can inspect this schema further by logging into the MySQL CLI.

You should see the following

```
Welcome to the MySQL monitor. Commands end with; or \g.
Your MySQL connection id is 1138
Server version: 5.7.27-log MySQL Community Server (GPL)

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Type 'help;' or '\h' for help. Type '\c' to clear the current input statement.

mysql>
```

To view your tables.

```
show tables;
```

There's an extra table here called dc01_out_of_stock_events that not in the schema diagram above, we'll cover this table separately later on.

...and get some row counts.

```
SELECT * from (

SELECT 'customers' as table_name, COUNT(*) FROM customers

UNION

SELECT 'products' as table_name, COUNT(*) FROM products

UNION

SELECT 'suppliers' as table_name, COUNT(*) FROM suppliers

UNION

SELECT 'sales_orders' as table_name, COUNT(*) FROM sales_orders

UNION

SELECT 'sales_order_details' as table_name, COUNT(*) FROM sales_order_details

UNION

SELECT 'purchase_orders' as table_name, COUNT(*) FROM purchase_orders

UNION

SELECT 'purchase_order_details' as table_name, COUNT(*) FROM purchase_orders

UNION

SELECT 'purchase_order_details' as table_name, COUNT(*) FROM purchase_order_details
) row_counts;
```

As you can see, we have 30 customers, suppliers and products, 0 sales orders and 1 purchase order.

table_name	-	UNT(*)
customers	+	30
products		30
suppliers		30
sales_orders		0
sales_order_details		0
purchase_orders		1
purchase_order_detai	ls	30

The single purchase order was created so we have something in stock to sell, let's have a look at what was ordered.

```
SELECT * FROM purchase_order_details;
```

	purchase_order_id	product_id		cost +
1		1		 6.82
2	1	2	100	7.52
3	1	3	100	6.16
4	1	4	100	8.07
5	1	5	100	2.10
6	1	6	100	7.45
7	1	7	100	4.02
8	1	8	100	0.64
9	1	9	100	8.51
10	1	10	100	3.61
11	1	11	100	2.62
12	1	12	100	2.60
13	1	13	100	1.26
14	1	14	100	4.08
15	1	15	100	3.56
16	1	16	100	7.13
17	1	17	100	7.64
18	1	18	100	5.94
19	1	19	100	2.94
20	1	20	100	1.91
21	1	21	100	8.89
22	1	22	100	7.62
23	1	23	100	6.19
24	1	24	100	2.83
25	1	25	100	5.51
26	1	26	100	4.23
27	1	27	100	8.33
28	1	28	100	7.09
29	1	29	100	1.75
30	1	30	100	1.72

as you can see, we have ordered 100 of each product, this reflects our initial and current stock levels.

Type exit to leave the MySQL CLI

Starting the Orders Application

To start generating some sales orders we need to start the orders application. This application will continuously create new sales orders to simulate product demand. The application will also raise purchase orders when told to do so, we'll cover this aspect later on in the workshop.

Start the orders application by running the following command.

```
docker exec -dit db-trans-simulator sh -c "python -u /simulate_dbtrans.py > /proc/1/fd/1"
```

Confirm that the simulator is working as expected

```
docker logs -f db-trans-simulator
```

You should see an output like this:

```
Sales Order 1 Created
Sales Order 2 Created
Sales Order 3 Created
Sales Order 4 Created
Sales Order 5 Created
Sales Order 6 Created
Sales Order 7 Created
Sales Order 8 Created
Sales Order 9 Created
...
```

Press ctr1-c to quit

We now have sales orders being automatically created for us.

To confirm this, start the MySQL CLI again

```
docker exec -it mysql bash -c 'mysql -u root -p$MYSQL_ROOT_PASSWORD --database orders'
```

Re-run the row count script multiple times to confirm that the number of sales orders and sales order detail row counts are increasing.

```
SELECT * from (

SELECT 'customers' as table_name, COUNT(*) FROM customers

UNION

SELECT 'products' as table_name, COUNT(*) FROM products

UNION

SELECT 'suppliers' as table_name, COUNT(*) FROM suppliers

UNION

SELECT 'sales_orders' as table_name, COUNT(*) FROM sales_orders

UNION

SELECT 'sales_order_details' as table_name, COUNT(*) FROM sales_order_details

UNION

SELECT 'purchase_orders' as table_name, COUNT(*) FROM purchase_orders

UNION

SELECT 'purchase_order_details' as table_name, COUNT(*) FROM purchase_orders

UNION

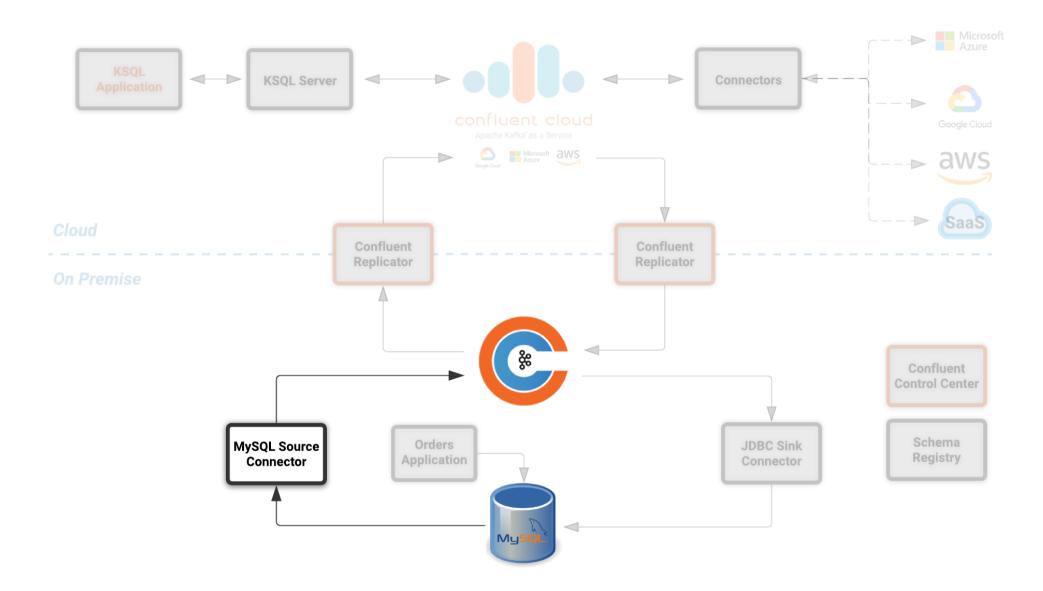
SELECT 'purchase_order_details' as table_name, COUNT(*) FROM purchase_order_details
) row_counts;
```

customers	1	30
products	- 1	30
suppliers	- 1	30
sales_orders	- 1	130
sales_order_details	- 1	392
purchase_orders	- 1	1
purchase_order_detail	.s	30

Type exit to leave the MySQL CLI

Lab 3: Stream CDC events to your Local Kafka Cluster

Now that we have data being automatically created in our MySQL database it's time to stream those changes into your onpremise Kafka cluster. We can do this using the Debezium MySQL Source connector



Create the MySQL source connector

We have a Kafka Connect worker already up and running in a docker container called kafka-connect-onprem. This Kafka Connect worker is configured to connect to your on-premise Kafka cluster and has a internal REST server listening on port 18083. We can create a connector from the command line using the cURL command. The cURL command allows us to send an HTTP POST request to the REST server, the '-H' option specifies the header of the request and includes the target host and port information, the -d option specifies the data we will send, in this case its the configuration options for the connector. You can of course create and manage connectors using any tool or language capable of issuing HTTP requests.

To create the Debezium MySQL Source connector instance run the following command:-

```
"database.user": "mysqluser",
        "database.password": "mysqlpw",
        "database.server.id": "12345",
        "database.server.name": "dc01",
        "database.whitelist": "orders",
        "table.blacklist": "orders.dc01_out_of_stock_events",
        "database.history.kafka.bootstrap.servers": "broker:29092",
        "database.history.kafka.topic": "debezium_dbhistory",
        "include.schema.changes": "true",
        "snapshot.mode": "when_needed",
        "transforms": "unwrap, sourcedc, TopicRename",
        "transforms.unwrap.type": "io.debezium.transforms.UnwrapFromEnvelope",
        "transforms.sourcedc.type":"org.apache.kafka.connect.transforms.InsertField$Value",
        "transforms.sourcedc.static.field": "sourcedc",
        "transforms.sourcedc.static.value":"dc01",
        "transforms.TopicRename.type": "org.apache.kafka.connect.transforms.RegexRouter",
        "transforms.TopicRename.regex": "(.*)\\.(.*)\\.(.*)",
        "transforms.TopicRename.replacement": "$1_$3"
    }
}'
```

The output should resemble something similar to this...

```
HTTP/1.1 201 Created

Date: Thu, 20 Feb 2020 13:00:57 GMT

Location: http://localhost:18083/connectors/mysql-source-connector

Content-Type: application/json

Content-Length: 1043

Server: Jetty(9.4.20.v20190813)

...

...
```

View Messages in Confluent Control Center

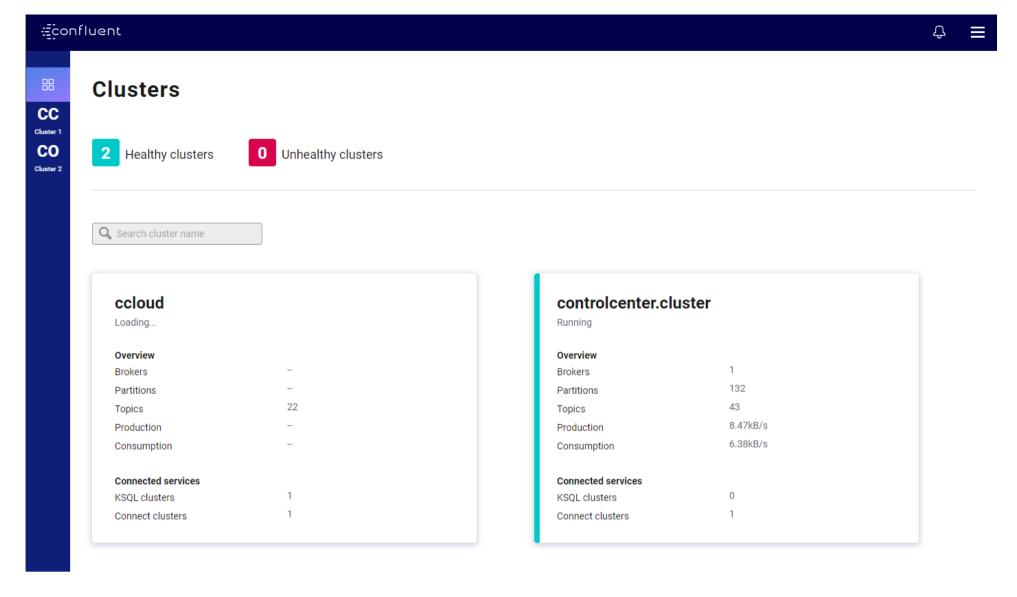
Now that the MySQL source connector is up and running, we will be able to see messages appear in our local Kafka cluster.

We can use Confluent Control Center to confirm this.

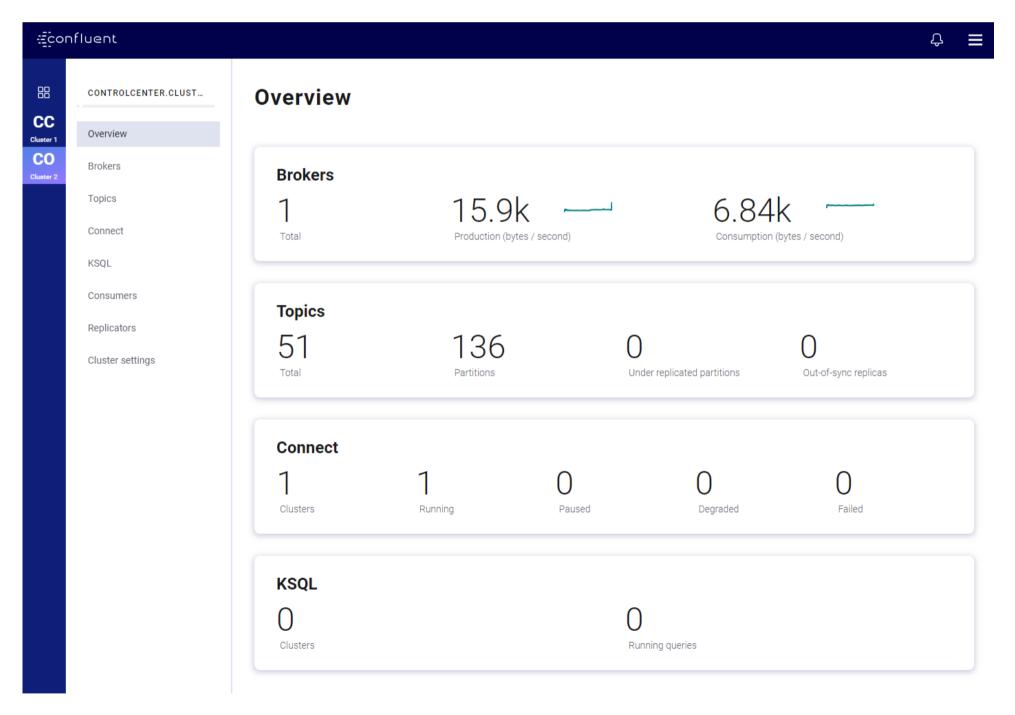
Use the following and username and password to authenticate to Confluent Control Center

Usernam Passwor	ne: dc01 rd:					
Sign in http://34.89 Your conne	9.9.90:9021 ction to this site is not private					
Username						
Password						
		Sign in Co	ancel			

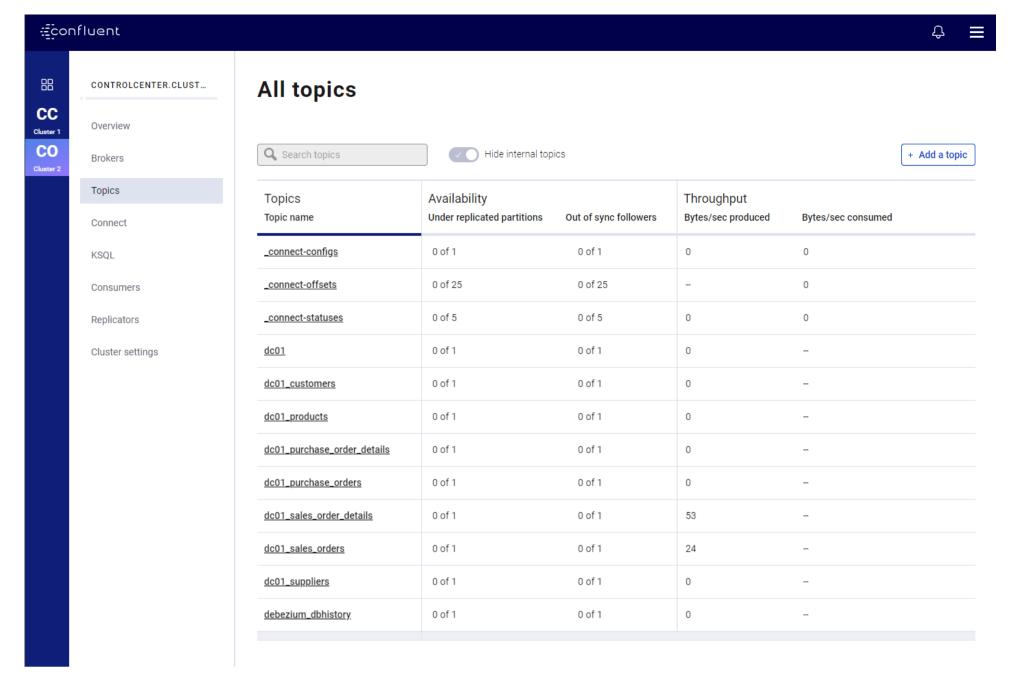
On the landing page we can see that Confluent Control Center is monitoring two Kafka Clusters, our on-premise cluster and a Confluent Cloud Cluster



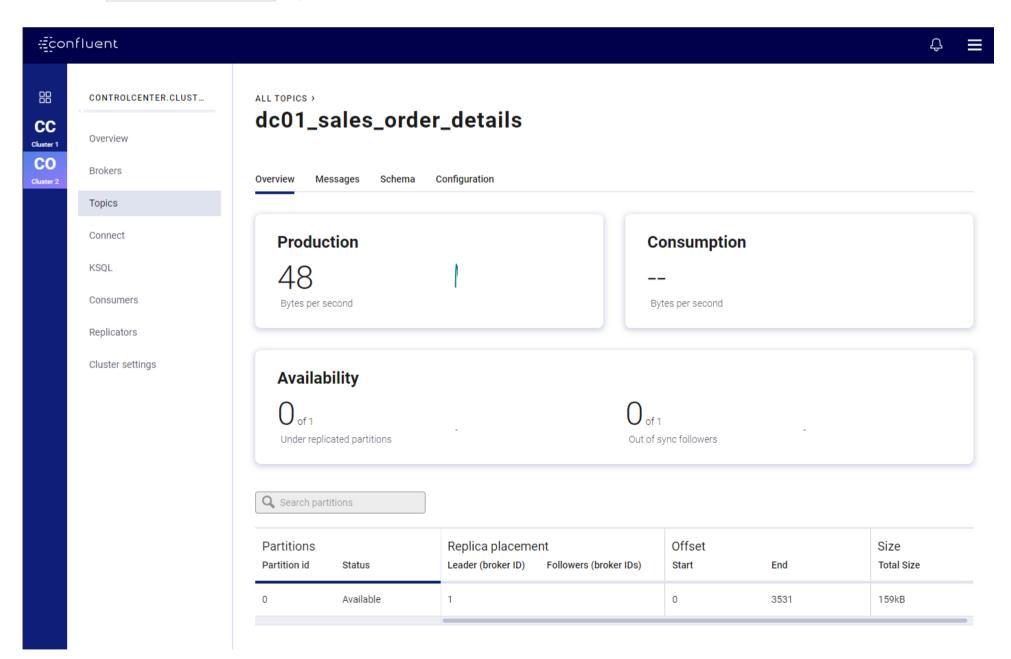
On the left hand navigation bar select "CO" (Controlcenter.cluster), this is your on-premise cluster.



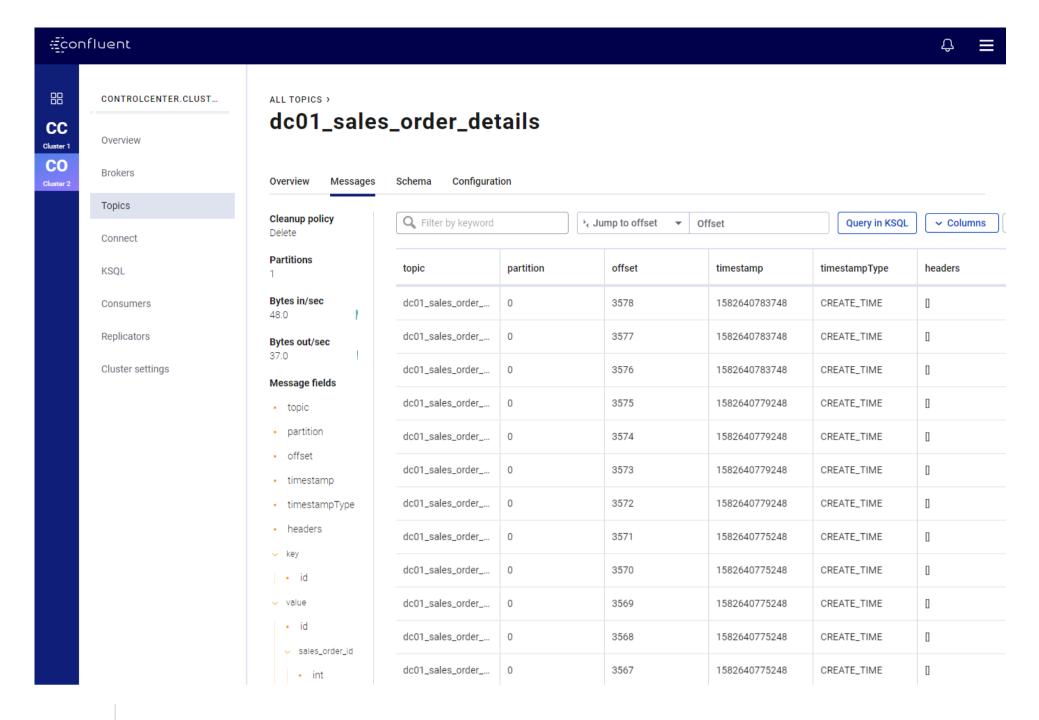
Select the Topics Menu on the left



Select the dc01_sales_order_details topic



Finally select the Messages tab and observe that messages are being streamed into Kafka from MySQL in real time.



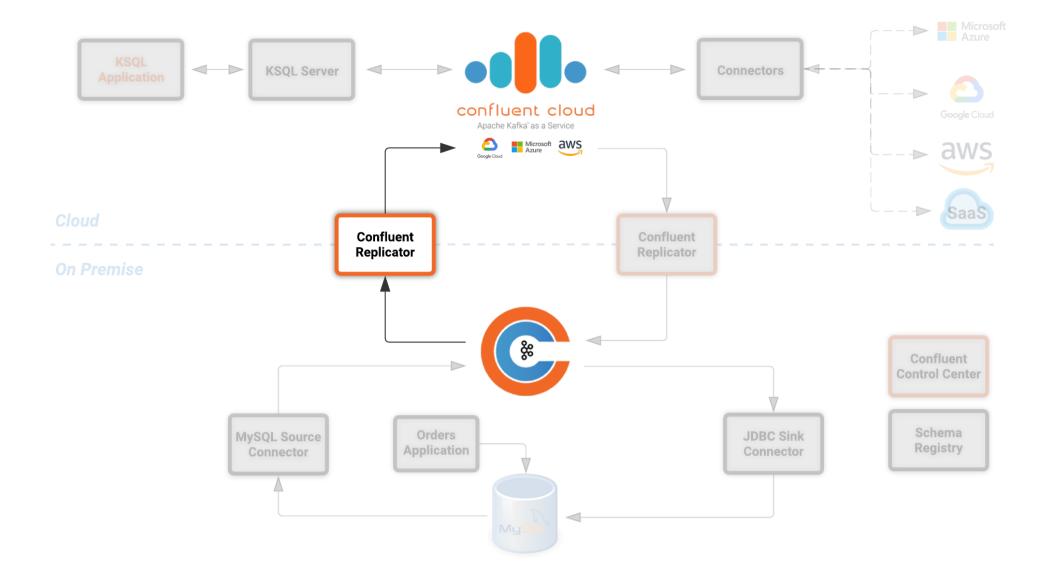
Further Reading



- Debezium MySQL Configuration Options
- Kafka Connect REST API
- cURL manpage
- Confluent Control Center Documentation

Lab 4: Stream events to Confluent Cloud

Now that your on-premise Kafka cluster is receiving events from your MySQL Database let's use Confluent Replicator to stream those messages to Confluent Cloud



Create the Replicator Connector Instance

Confluent Replicator uses Kafka Connect under the covers and can be considered a special type of connector, however, unlike other connectors, the source *and* target technology for the connector is a Kafka Cluster.

To support this connector, we have another Kafka Connect worker running in a different docker container called kafka-connect-ccloud. This Kafka Connect worker is configured to connect to the Confluent Cloud instance provisioned for this workshop.

This Kafka Connect worker has an internal REST server listening on port 18084.

Run the following from the command line to create the Replicator Connector instance, this connector will replicate events from you on-premise Kafka cluster to your Confluent Cloud Cluster.

```
curl -i -X POST -H "Accept:application/json" \
    -H "Content-Type:application/json" http://localhost:18084/connectors/ \
    -d '{
        "name": "replicator-dc01-to-ccloud",
        "config": {
          "connector.class": "io.confluent.connect.replicator.ReplicatorSourceConnector",
          "key.converter": "io.confluent.connect.replicator.util.ByteArrayConverter",
          "value.converter": "io.confluent.connect.replicator.util.ByteArrayConverter",
          "topic.config.sync": false,
          "topic.regex": "dc[0-9][0-9][_].*",
          "topic.blacklist": "dc01_out_of_stock_events",
          "dest.kafka.bootstrap.servers": "${file:/secrets.properties:CCLOUD_CLUSTER_ENDPOINT}",
          "dest.kafka.security.protocol": "SASL_SSL",
          "dest.kafka.sasl.mechanism": "PLAIN",
          "dest.kafka.sasl.jaas.config": "org.apache.kafka.common.security.plain.PlainLoginModule required
username=\"${file:/secrets.properties:CCLOUD_API_KEY}\" password=\"${file:/secrets.properties:CCLOUD_API_SECRET}\";",
          "dest.kafka.replication.factor": 3,
          "src.kafka.bootstrap.servers": "broker:29092",
          "src.consumer.group.id": "replicator-dc01-to-ccloud",
          "src.consumer.interceptor.classes":
"io.confluent.monitoring.clients.interceptor.MonitoringConsumerInterceptor",
          "src.consumer.confluent.monitoring.interceptor.bootstrap.servers": "broker:29092",
          "src.kafka.timestamps.producer.interceptor.classes":
"io.confluent.monitoring.clients.interceptor.MonitoringProducerInterceptor",
```

```
"src.kafka.timestamps.producer.confluent.monitoring.interceptor.bootstrap.servers": "broker:29092",
    "tasks.max": "1"
}
```

You should see something similar...

```
HTTP/1.1 100 Continue

HTTP/1.1 201 Created

Date: Sun, 09 Feb 2020 15:07:22 GMT

Location: http://localhost:18084/connectors/replicator-dc01-to-ccloud

Content-Type: application/json

Content-Length: 1342

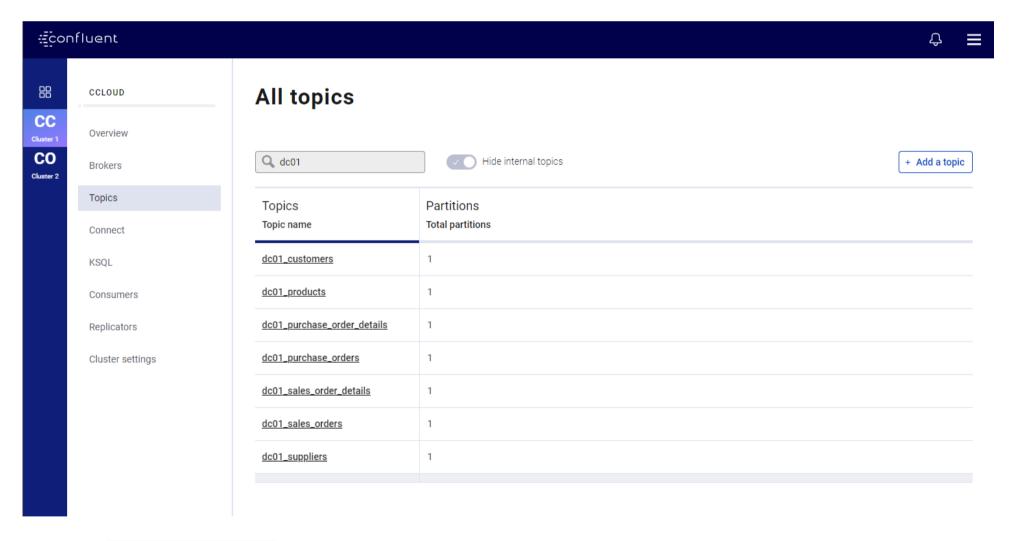
Server: Jetty(9.4.20.v20190813)
...
...
```

Confirm that Messages are Arriving in Confluent Cloud

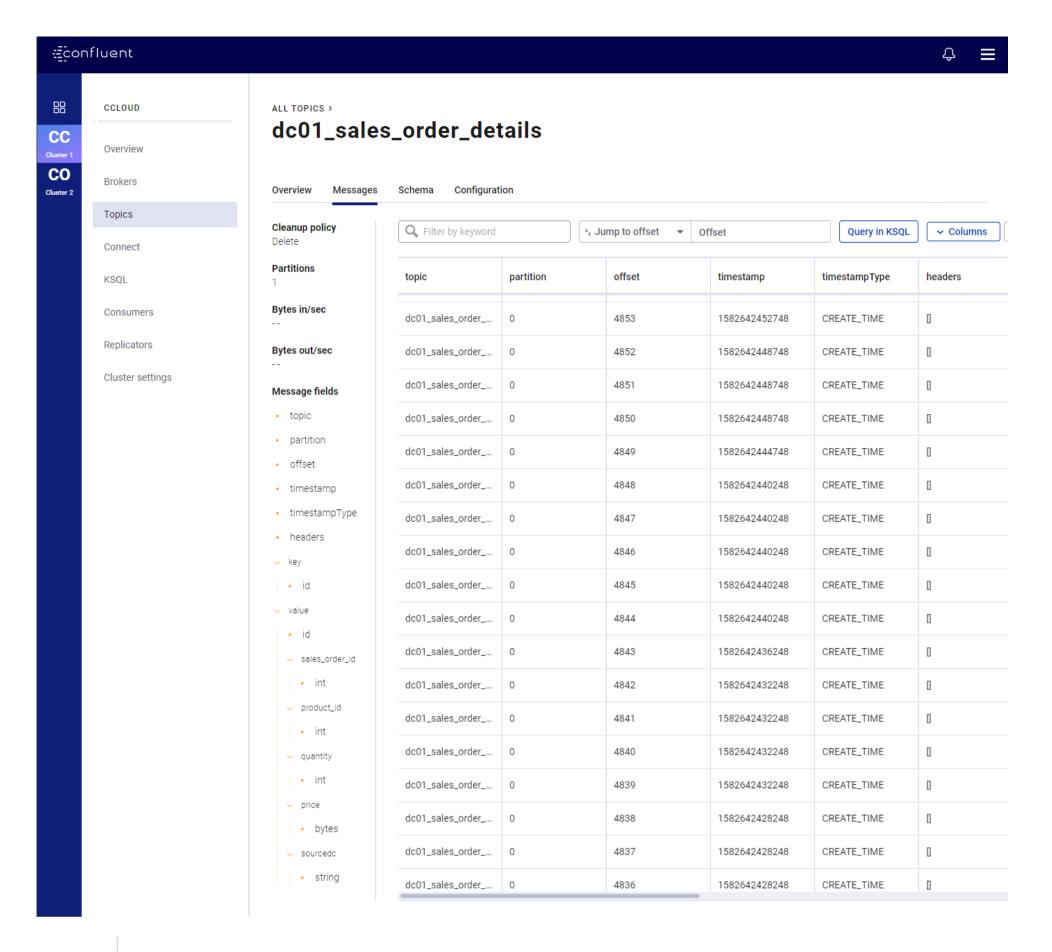
Jump back to Confluent Control Center

Select the "CC" cluster from the left-hand navigation bar and then select "Topics".

This Confluent Cloud Instance is being shared by other users of the workshop and as a result you will see topics being replicated from other data centers. To see just your topics, type your data center name, dcO1, into the search box at the top to filter.



Select the dc01_sales_order_details topic and finally the "Messages" tab under the topic heading. You should see messages streaming in from you on-premise Kafka cluster.



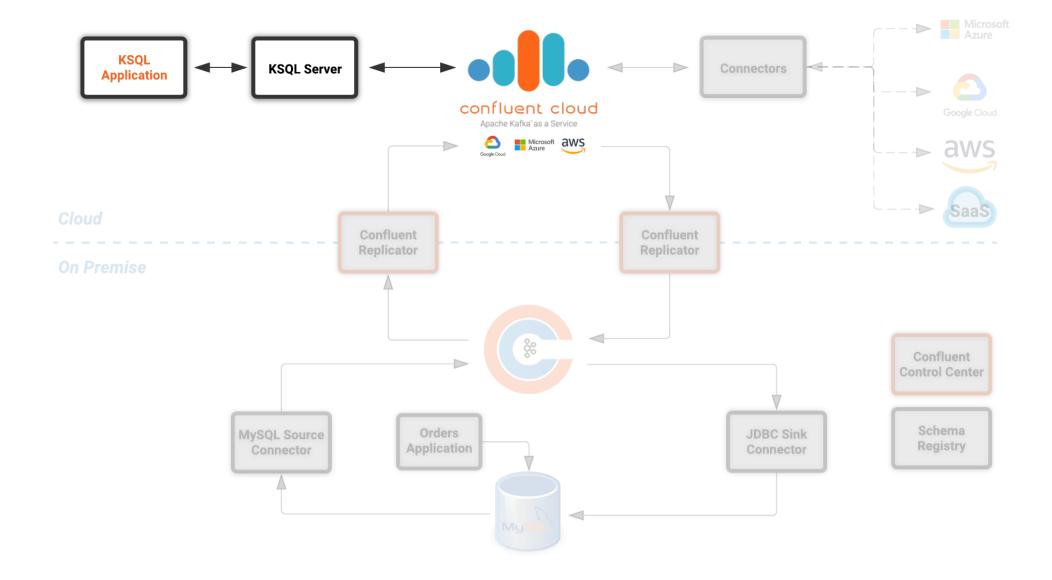


Further Reading

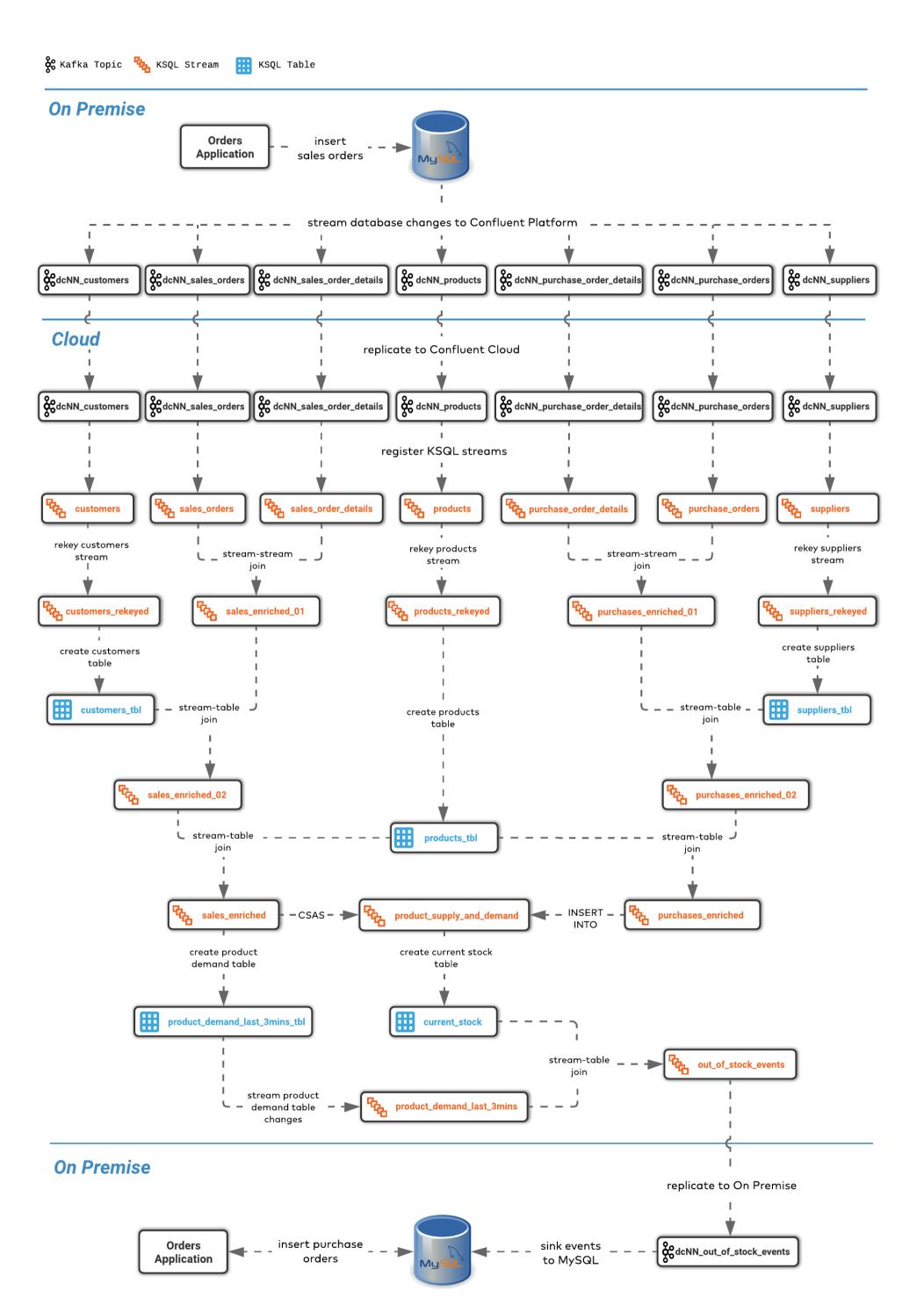
- Confluent Replicator
- Confluent Replicator Configuration Properties

Lab 5: Building a Realtime Supply & Demand Application using KSQL

We now have all the data we need being streamed, in real time, to Confluent Cloud. You have a KSQL Server running inside a docker container that is configured to point to our Confluent Cloud cluster. In a real world deployment, it is likely that this KSQL Server would be running closer to Confluent Cloud but for the purposes of this workshop it is not important.



Below is an illustration of the completed Supply & Demand KSQL Application, over the next few labs you will be building this step-by-step.



Lab 6: Getting Started with KSQL

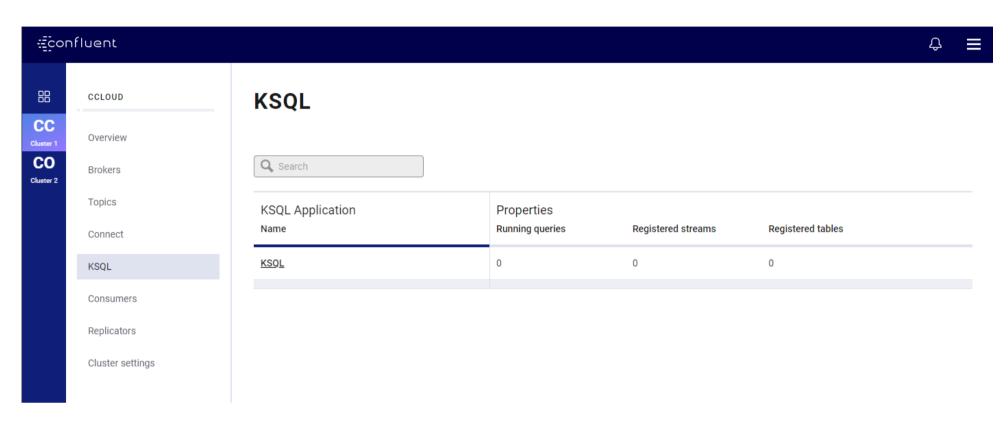
You can interact with KSQL Server using either the KSQL CLI, Confluent Control Center or the REST API. This workshop will focus on the KSQL CLI but if you'd rather use Confluent Control Center then read the next section.

Using KSQL with Confluent Control Center

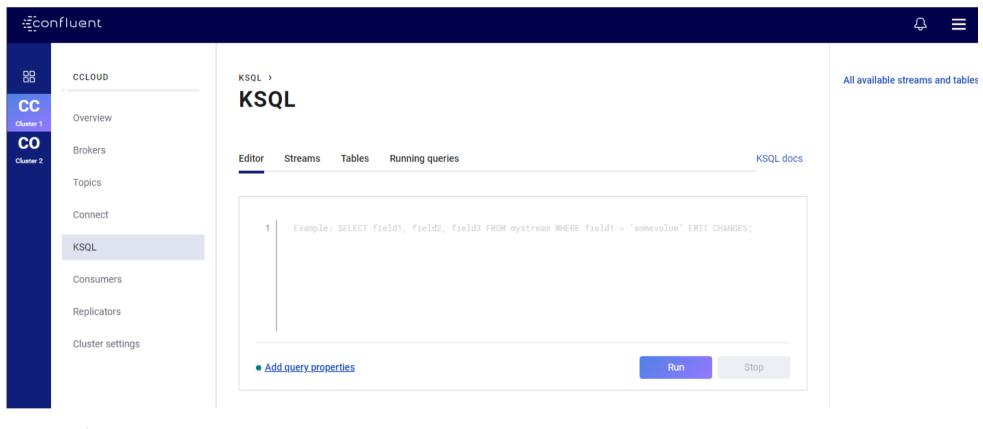
If you'd rather use Confluent Control Center then follow the instructions below, otherwise skip this section.

Open Confluent Control Center

Click the "CC" Cluster on the left-hand navigation bar, Select "KSQL" and finally click on the "KSQL" application.



You will now be able to use the "Editor" tab instead of the CLI





This workshop will focus on the KSQL CLI

Start the KSQL CLI

To start the KSQL CLI run the following command:-

docker exec -it ksql-cli ksql http://ksql-server-ccloud:8088

You should see something like this:-

The KSQL CLI is pointing at a KSQL Server connected to your Confluent Cloud instance.

To view a list of all topics in Confluent Cloud run the following command:-

```
show topics;
```

You should see your own topics, dc01_*, along with topics from other workshop users.

```
ksql> show topics;
                             | Partitions | Partition Replicas
Kafka Topic
 _confluent-command
                             | 1
                                          | 3
                                          | 3
                             | 1
 _dc01-connect-configs
                                          | 3
 _dc01-connect-offsets
                             | 1
                             | 1
                                          | 3
 _dc01-connect-statuses
 _dc02-connect-configs
                             | 1
                                          | 3
                                          | 3
 _dc02-connect-offsets
                             | 1
                                          | 3
 _dc02-connect-statuses
                             | 1
                             | 1
                                          | 3
 dc01_customers
                                          | 3
dc01_products
                             | 1
 dc01_purchase_order_details | 1
                                          | 3
                                          | 3
 dc01_purchase_orders
                             | 1
                                          | 3
dc01_sales_order_details
                             | 1
 dc01_sales_orders
                             | 1
                                          | 3
                             | 1
                                          | 3
 dc01_suppliers
 dc02 customers
                             | 1
                                          | 3
 dc02_products
 dc02_purchase_order_details | 1
                                          | 3
 dc02_purchase_orders
                                          | 3
dc02_sales_order_details
                             | 1
                                          | 3
. . .
```

Inspect a topic's contents

To inspect the contents of a topic run the following:-

```
PRINT dc01_sales_orders;
```

You should see something similar:-

```
ksql> PRINT dc01_sales_orders;
Format:AVRO
2/20/20 1:23:55 PM UTC,
•, {"id": 466, "order_date": 1582205036000, "customer_id": 12, "sourcedc": "dc01"}
2/20/20 1:23:59 PM UTC,
•, {"id": 467, "order_date": 1582205040000, "customer_id": 27, "sourcedc": "dc01"}
2/20/20 1:24:03 PM UTC,
•, {"id": 468, "order_date": 1582205044000, "customer_id": 20, "sourcedc": "dc01"}
2/20/20 1:24:07 PM UTC,
•, {"id": 469, "order_date": 1582205048000, "customer_id": 7, "sourcedc": "dc01"}
2/20/20 1:24:11 PM UTC,
•, {"id": 470, "order_date": 1582205052000, "customer_id": 30, "sourcedc": "dc01"}
2/20/20 1:24:15 PM UTC,
•, {"id": 471, "order_date": 1582205056000, "customer_id": 27, "sourcedc": "dc01"}
2/20/20 1:24:20 PM UTC,
•, {"id": 472, "order_date": 1582205060000, "customer_id": 8, "sourcedc": "dc01"}
2/20/20 1:24:24 PM UTC,
•, {"id": 473, "order_date": 1582205064000, "customer_id": 8, "sourcedc": "dc01"}
```

Press ctrl-c to stop



The events streaming from the MySQL database are serialized with Avro and as a result you will see some special characters in the above output, this is because the "PRINT TOPIC" command uses the String description.

KSQL Streams

In order to work with a stream of data in KSQL we first need to register a KSQL Stream over an existing topic.

We can do this using a CREATE STREAM statement. Run the following command to create your first KSQL stream:-

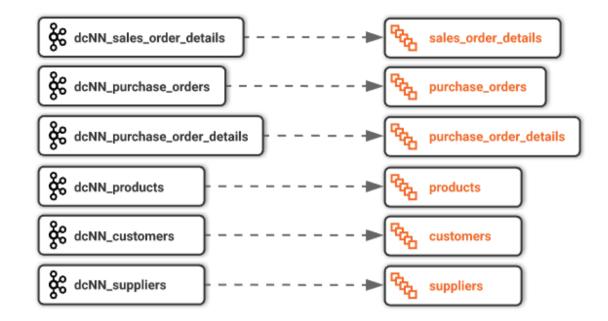
You should see the following output

```
ksql> CREATE STREAM sales_orders WITH (KAFKA_TOPIC='dc01_sales_orders', VALUE_FORMAT='AVRO');

Message
------
Stream created
------
```

Create streams for each of your remaining topics

```
CREATE STREAM sales_order_details WITH (KAFKA_TOPIC='dc01_sales_order_details', VALUE_FORMAT='AVRO');
CREATE STREAM purchase_orders WITH (KAFKA_TOPIC='dc01_purchase_orders', VALUE_FORMAT='AVRO');
CREATE STREAM purchase_order_details WITH (KAFKA_TOPIC='dc01_purchase_order_details', VALUE_FORMAT='AVRO');
CREATE STREAM products WITH (KAFKA_TOPIC='dc01_products', VALUE_FORMAT='AVRO');
CREATE STREAM customers WITH (KAFKA_TOPIC='dc01_customers', VALUE_FORMAT='AVRO');
CREATE STREAM suppliers WITH (KAFKA_TOPIC='dc01_suppliers', VALUE_FORMAT='AVRO');
```



To view your current streams run the following command:-

```
SHOW STREAMS;
```

Notice that each stream is mapped to an underlying Kafka topic and that the format is AVRO.

Stream Name	Kafka Topic	Format
CUSTOMERS	dc01_customers	AVRO
PRODUCTS	dc01_products	AVRO
PURCHASE_ORDERS	dc01_purchase_orders	AVRO
PURCHASE_ORDER_DETAILS	dc01_purchase_order_details	AVRO
SALES_ORDERS	dc01_sales_orders	AVRO
SALES_ORDER_DETAILS	dc01_sales_order_details	AVRO
SUPPLIERS	dc01_suppliers	AVRO

To view the details of an individual topic you can you can use the describe command:-

```
DESCRIBE sales_order_details;
```

Notice that all the columns have been created for us and we didn't need to explicitly set their names and data types when we created the stream, this is one of the advantages of using AVRO and the Schema Registry.

Also notice that KSQL adds the implicit columns ROWTIME and ROWKEY to every stream and table, which represent the corresponding Kafka message timestamp and message key, respectively. The timestamp has milliseconds accuracy.

```
: SALES_ORDER_DETAILS
Name
Field
               | Type
ROWTIME
               BIGINT
                                  (system)
ROWKEY
               | VARCHAR(STRING) (system)
ID
               INTEGER
SALES_ORDER_ID | INTEGER
PRODUCT_ID
               INTEGER
QUANTITY
               INTEGER
               DECIMAL
PRICE
               | VARCHAR(STRING)
SOURCEDC
For runtime statistics and query details run: DESCRIBE EXTENDED;
```

Further Reading



- KSQL Overview
- KSQL Streams
- CREATE STREAM Syntax

Lab 7: Querying Streams using KSQL

There are two types of query in KSQL, **Push** queries and **Pull** queries.

- Push Queries enable you to subscribe to a result as it changes in real-time. You can subscribe to the output of any query, including those that return a stream or a materialized aggregate table. The EMIT CHANGES clause is used to indicate a query is a push query.
- Pull Queries are a preview feature with KSQL 5.4 and enable you to look up information at a point in time.

Another important point to understand is where within a topic a query starts to read from. You can control this behaviour using the ksql.streams.auto.offset.reset property. This property can either be set to earliest where data is consumed from the very beginning of the topic or latest where only new data is consumed.

To see the current values for all properties run the following command

```
SHOW PROPERTIES;
```

Look out for a property called ksq1.streams.auto.offset.reset, it should be set to latest as this is the default setting configured on the KSQL server.

```
Property | Default override | Effective Value
....
ksql.streams.auto.offset.reset | latest
...
ksql>
```

You can override this setting to suit you needs:-

```
SET 'ksql.streams.auto.offset.reset'='earliest';
SET 'ksql.streams.auto.offset.reset'='latest';
```

Or preferably, using the abbreviated property names:-

```
SET 'auto.offset.reset' = 'latest';
SET 'auto.offset.reset' = 'earliest';
```

Let's start by running a Push guery and consume all messages from the beginning of a stream.

You should see something similar to this:-

ID	SALES_ORDER_ID	PRODUCT_ID	QUANTITY	PRICE	1
+ 1	+ 1	1	+ 10	+ 2.68	+·
2	1	23	1	9.01	ĺ
3	1	14	6	5.84	I
4	2	12	7	4.00	I
5	2	9	4	9.83	
6	2	5	1	8.81	
7	2	3	8	9.99	
8	2	1	9	2.68	
9	3	21	5	9.90	
10	3	2	1	8.23	I
11	3	4	2	9.78	1
12	4	15	2	6.16	1
•••					
•••					
480	157	26	5	9.03	1
481	158	2	2	8.23	- 1
482	159	10	4	5.32	- 1
483	160	25	8	9.00	!

Press ctrl-c to stop

Notice that events continue to stream to the console until you explicitly cancel the query, this is because when we are working with streams in KSQL the data set is unbounded and could theoretically continue forever.

To inspect a bounded set of data, you can use the LIMIT clause.

Here we are seeing the first 10 messages that were written to the topic. Notice that the query automatically terminates when the limit of 10 events is reached.

ID	ALES_ORDER_ID	PRODUCT_ID	QUANTITY	PRICE	
1 1		1	10	2.68	
2 1		23	1	9.01	١
3 1		14	6	5.84	I
4 2		12	7	4.00	
5 2		9	4	9.83	
6 2		5	1	8.81	
7 2		3	8	9.99	
8 2		1	9	2.68	
9 3		21	5	9.90	
10 3		2	1	8.23	
Limit Reached					
Query terminated					
ksql>					

Filtering Streams

Since KSQL is based on SQL, you can do many of the standard SQL things you'd expect to be able to do, including predicates and projections. The following query will return a stream of you the latest sales orders where the quantity column is greater than 3.

```
SET 'auto.offset.reset'='latest';
SELECT id,
          product_id,
          quantity
FROM sales_order_details
WHERE quantity > 3
EMIT CHANGES;
```

You should only see events where the quantity column value is greater than 3.

ID	PRODUCT_ID	QUANTITY	1
+			+
3153	22	8	1
3154	4	6	
3155	9	4	1
3156	25	10	1
3158	24	8	1
3159	7	4	1
3161	28	8	1
3162	22	7	1
3163	24	 6	
3165	 5	8	1
3167	21	9	1

Press ctrl-c to stop



Further Reading

- Push Query Syntax
- Pull Query Syntax
- KSQL Offset Management

Lab 8: Creating KSQL tables

KSQL tables allow you to work the data in topics as key/value pairs, with a single value for each key. KSQL tables can be created from an existing topic or from the query results from other tables or streams. You can read more about this here.

We want to create tables over the <code>customers</code>, <code>suppliers</code> and <code>products</code> streams so we can look up the current state for each customer, supplier and product. Later in the workshop we will want to join these tables to other streams. To successfully join to a table in KSQL you need to ensure that the table is keyed on the column you are going to use in the join. To achieve this, we need to make sure the stream that we are creating a table from is keyed correctly.

Rekeying Streams

We can see what the current key for stream or table is by using the DESCRIBE EXTENDED command.

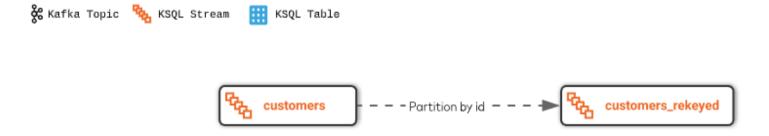
```
DESCRIBE EXTENDED customers;
```

You can see in the output that the Key Field is not set.

Name : CUSTOMERS Type : STREAM Key field : STRING Key format Timestamp field : Not set - using Value format : AVRO Kafka topic : dc01_customers (partitions: 1, replication: 3) Field Type ROWTIME BIGINT (system) ROWKEY | VARCHAR(STRING) (system) INTEGER ID FIRST_NAME | VARCHAR(STRING) LAST_NAME | VARCHAR(STRING) EMAIL | VARCHAR(STRING) CITY | VARCHAR(STRING) COUNTRY | VARCHAR(STRING) | VARCHAR(STRING) SOURCEDC

We can fix this by creating a derived stream that has the correct key.

```
SET 'auto.offset.reset'='earliest';
CREATE STREAM customers_rekeyed WITH (KAFKA_TOPIC='dc01_customers_rekeyed', PARTITIONS=1) AS
SELECT * FROM customers
PARTITION BY id;
```



This method of creating a derived topic is frequently referred to by the acronym $(SAS) \rightarrow (CREATE STREAM ... AS SELECT)$ where we create a new topic based on the contents of another. Unlike CSAS statements in a traditional RDBMS, CSAS statements in KSQL create *continuous queries* where data is continuously streamed from the source topic into the target topic.

We can confirm that the new stream has the correct key by running the DESCRIBE EXTENDED command again

```
DESCRIBE EXTENDED customers_rekeyed;
```

You can see in the output that the Key Field is now set correctly.

```
: CUSTOMERS_REKEYED
Name
Type
                     : STREAM
Key field
                     : CUSTOMERS_REKEYED.ID
Key format
                     : STRING
Timestamp field
                     : Not set - using
Value format
                     : AVRO
                     : dc01_customers_rekeyed (partitions: 1, replication: 3)
Kafka topic
Field
            | Type
ROWTIME
            BIGINT
                               (system)
            | VARCHAR(STRING) (system)
ROWKEY
```

```
ID
           INTEGER
                             (key)
FIRST_NAME | VARCHAR(STRING)
LAST_NAME | VARCHAR(STRING)
EMAIL
           | VARCHAR(STRING)
CITY
           | VARCHAR(STRING)
COUNTRY
           | VARCHAR(STRING)
          | VARCHAR(STRING)
SOURCEDC
Queries that write from this STREAM
_____
CSAS_CUSTOMERS_REKEYED_10 : CREATE STREAM CUSTOMERS_REKEYED WITH (KAFKA_TOPIC='dc01_customers_rekeyed', PARTITIONS=1,
REPLICAS=3) AS SELECT *
FROM CUSTOMERS CUSTOMERS
EMIT CHANGES
PARTITION BY ID;
For query topology and execution plan please run: EXPLAIN
Local runtime statistics
______
messages-per-sec: 0.30 total-messages:
                                                 30
                                                        last-message: 2020-02-26T12:11:31.227Z
(Statistics of the local KSQL server interaction with the Kafka topic dc01_customers_rekeyed)
```

In the above output also notice the Queries that write from this STREAM section, here you can see the query you just ran, this a called a persistant query and runs in the background continuously streaming messages until it is terminated.

You can view the current persistant queries that are running using the following command:-

```
Query ID | Kafka Topic | Query String

CSAS_CUSTOMERS_REKEYED_10 | CUSTOMERS_REKEYED | CREATE STREAM CUSTOMERS_REKEYED WITH

(KAFKA_TOPIC='dc01_customers_rekeyed', PARTITIONS=1, REPLICAS=3) AS SELECT *

FROM CUSTOMERS CUSTOMERS

EMIT CHANGES

PARTITION BY ID;

For detailed information on a Query run: EXPLAIN;
```

Now that we have our customers_rekeyed stream created let's rekey the supplier and products streams.

```
SET 'auto.offset.reset'='earliest';

CREATE STREAM products_rekeyed WITH (KAFKA_TOPIC='dc01_products_rekeyed', PARTITIONS=1) AS

SELECT * FROM products

PARTITION BY id;

CREATE STREAM suppliers_rekeyed WITH (KAFKA_TOPIC='dc01_suppliers_rekeyed', PARTITIONS=1) AS

SELECT * FROM suppliers

PARTITION BY id;
```



Another way to confirm a stream is keyed correctly, is to query it and compare the system ROWKEY column to your join column, they should be identical.

```
SELECT rowkey, id FROM customers_rekeyed EMIT CHANGES LIMIT 3;
SELECT rowkey, id FROM products_rekeyed EMIT CHANGES LIMIT 3;
SELECT rowkey, id FROM suppliers_rekeyed EMIT CHANGES LIMIT 3;
```

You can see in the output that the ROWKEY and ID columns have identical values for each stream.

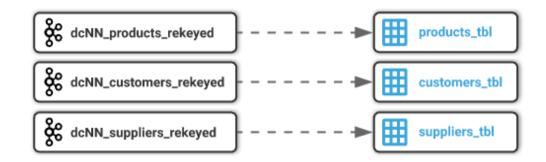
ROWKEY	ID	I
+ 1	1	
2	2	İ
3	3	

ROWKEY	ID	1
+ 1	1	+
2	2	ĺ
3	3	

Creating Tables

We are now in a position where we can create our first KSQL tables. To do this we need to register tables with KSQL over the newly re-keyed topics.

```
CREATE TABLE customers_tbl WITH (KAFKA_TOPIC='dc01_customers_rekeyed', VALUE_FORMAT='AVRO', key='id');
CREATE TABLE products_tbl WITH (KAFKA_TOPIC='dc01_products_rekeyed', VALUE_FORMAT='AVRO', key='id');
CREATE TABLE suppliers_tbl WITH (KAFKA_TOPIC='dc01_suppliers_rekeyed', VALUE_FORMAT='AVRO', key='id');
```



We can view our current tables using the following command:-

```
Table Name | Kafka Topic | Format | Windowed

CUSTOMERS_TBL | dc01_customers_rekeyed | AVRO | false
PRODUCTS_TBL | dc01_products_rekeyed | AVRO | false
SUPPLIERS_TBL | dc01_suppliers_rekeyed | AVRO | false
```

We'll use these tables soon and join them to our streams.

Further Reading



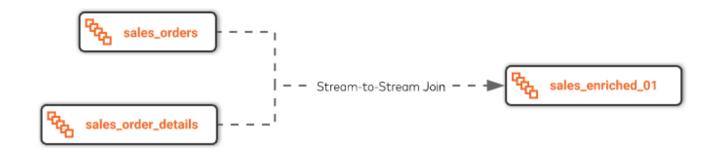
- CREATE TABLE Syntax
- DESCRIBE Syntax
- CREATE STREAM AS SELECT Syntax
- Message Key Requirements

Lab 9: KSQL Stream-to-Stream Joins

We can join two streams together in KSQL using a windowed join. When using a windowed join, you must specify a windowing scheme by using the WITHIN clause. A new input record on one side produces a join output for each matching record on the other side, and there can be multiple such matching records within a join window.

In the example below you can see that we have specified a window of 1 seconds using the WITHIN clause. The source application creates sales orders and their associated sales order detail rows at the same time, so a second will be plenty of time to ensure that a join takes place.

```
CREATE STREAM sales_enriched_01 WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_sales_enriched_01') AS SELECT
    o.id order_id,
    od.id order_details_id,
    o.order_date,
    o.customer_id,
    od.product_id,
    od.quantity,
    od.price
FROM sales_orders o
INNER JOIN sales_order_details od WITHIN 1 SECONDS ON (o.id = od.sales_order_id);
```



If we query this new stream...

...we can see that we have combined the data from both the sales_order and sales_order_details streams.

O_ID	OD_ID	ORDER_DATE	CUSTOMER_	ID PRODUCT_ID	QUANTITY	PRICE	- 1
+	+	+	-+	+	+	-+	+
1	1	28-02-20	23	21	2	9.90	- 1
1	2	28-02-20	23	14	10	5.84	
1	3	28-02-20	23	9	10	9.83	
2	4	28-02-20	20	19	3	3.38	
2	5	28-02-20	20	12	6	4.00	
2	6	28-02-20	20	6	6	8.24	
2	7	28-02-20	20	15	5	6.16	
2	8	28-02-20	20	22	10	8.19	
3	9	28-02-20	9	11	3	4.65	
4	10	28-02-20	12	20	6	4.86	- 1
Limit Rea	ched						



Further Reading

• Stream-Stream Joins

Lab 10: KSQL Stream-to-Table Joins

We can take this a step further by joining this new stream to a couple of the KSQL tables we created earlier.

To do this we'll need to create a new stream, sales_enriched_02, that'll stream the result of joining the sales_enriched_01 stream to the customers_tbl table.

```
CREATE STREAM sales_enriched_02 WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_sales_enriched_02') AS SELECT
    se.order_id,
    se.order_details_id,
    se.order_date,
    se.customer_id,
```

```
se.product_id,
se.quantity,
se.price,
ct.first_name,
ct.last_name,
ct.email,
ct.city,
ct.country
FROM sales_enriched_01 se
INNER JOIN customers_tbl ct ON (se.customer_id = ct.id);
```

```
KSQL Stream
IIII KSQL Table

| Continue of the contin
```

And last but not least we can join to our products table by creating our final stream sales_enriched which will be the result of joining the sales_enriched_02 stream to the products_tbl table.

```
CREATE STREAM sales_enriched WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_sales_enriched') AS SELECT
    se.order_id,
    se.order_details_id,
    se.order_date,
    se.product_id product_id,
    pt.name product_name,
    pt.description product_desc,
    se.price product_price,
    se.quantity product_qty,
    se.customer_id customer_id,
    se.first_name customer_fname,
    se.last_name customer_lname,
    se.email customer_email,
    se.city customer_city,
    se.country customer_country
FROM sales_enriched_02 se
INNER JOIN products_tbl pt ON (se.product_id = pt.id);
Kafka Topic 🍖 KSQL Stream
                            KSQL Table
                       sales_enriched_02
```



If we run a describe on this stream...

```
DESCRIBE sales_enriched;
```

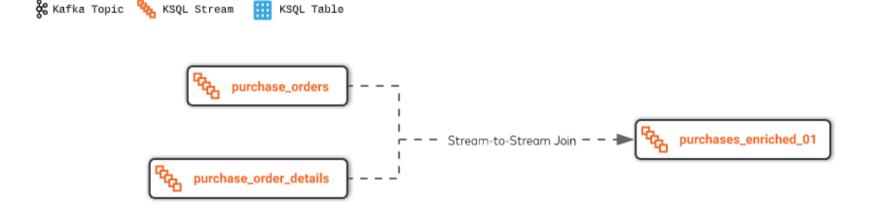
...you'll see that we have effectively denormalized the sales_orders, sales_order_details, customers and products streams/tables into a single event stream.

```
Field
                 | Type
                 BIGINT
                                    (system)
ROWTIME
ROWKEY
                  VARCHAR(STRING) (system)
ORDER_ID
                  INTEGER
ORDER_DETAILS_ID | INTEGER
ORDER_DATE
                 BIGINT
PRODUCT_ID
                  INTEGER
PRODUCT_NAME
                  VARCHAR(STRING)
PRODUCT_DESC
                 | VARCHAR(STRING)
PRODUCT_PRICE
                 DECIMAL
PRODUCT_QTY
                 INTEGER
CUSTOMER_ID
                 INTEGER
CUSTOMER_FNAME
                 | VARCHAR(STRING)
                  VARCHAR(STRING)
CUSTOMER_LNAME
CUSTOMER_EMAIL
                  VARCHAR(STRING)
CUSTOMER_CITY
                  VARCHAR(STRING)
CUSTOMER_COUNTRY | VARCHAR(STRING)
```

We now need to create an equivilent purchases_enriched stream that combines the purchase_orders, purchase_order_details, suppliers and products streams/tables. Since the purchases data model is very similar to that of the sales data model the process is the same.

Join the purchase_orders stream to the purchase_order_details stream

```
CREATE STREAM purchases_enriched_01 WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_purchases_enriched_01') AS SELECT
    o.id order_id,
    od.id order_details_id,
    o.order_date,
    o.supplier_id,
    od.product_id,
    od.quantity,
    od.cost
FROM purchase_orders o
INNER JOIN purchase_order_details od WITHIN 1 SECONDS ON (o.id = od.purchase_order_id);
```



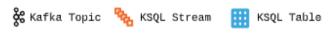
If we query this new stream...

...we can see that we have combined the data from both the purchase_order and purchase_order_details streams.

O_ID	OD_ID	ORDER_DATE	SUPPLIER_ID	PRODUCT_ID	QUANTITY	COST	
+	+	+	-+	-+	-+	+	+
1	1	02-03-20	1	1	100	6.82	
1	2	02-03-20	1	2	100	7.52	
1	3	02-03-20	1	3	100	6.16	- 1
1	4	02-03-20	1	4	100	8.07	- 1
1	5	02-03-20	1	5	100	2.10	
1	6	02-03-20	1	6	100	7.45	
1	7	02-03-20	1	7	100	4.02	- 1
1	8	02-03-20	1	8	100	0.64	- 1
1	9	02-03-20	1	9	100	8.51	1
1	10	02-03-20	1	10	100	3.61	1
Limit Rea	ched						
Query ter	minated						

Join the purchases_enriched_01 stream to the suppliers_tbl table...

```
CREATE STREAM purchases_enriched_02 WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_purchases_enriched_02') AS SELECT
    pe.order_id,
    pe.order_details_id,
    pe.order_date,
    pe.supplier_id,
    pe.product_id,
    pe.quantity,
    pe.cost,
    st.name,
    st.email,
    st.city,
    st.country
FROM purchases_enriched_01 pe
INNER JOIN suppliers_tbl st ON (pe.supplier_id = st.id);
```





...and finally join to the <code>products_tbl</code> table

```
CREATE STREAM purchases_enriched WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_purchases_enriched') AS SELECT
    pe.order_id,
    pe.order_details_id,
    pe.order_date,
    pe.product_id product_id,
    pt.name product_name,
    pt.description product_desc,
    pe.cost product_cost,
    pe.quantity product_qty,
    pe.supplier_id supplier_id,
    pe.name supplier_name,
```

```
pe.email supplier_email,
  pe.city supplier_city,
  pe.country supplier_country

FROM purchases_enriched_02 pe
INNER JOIN products_tbl pt ON (pe.product_id = pt.id);
```

If we run a describe on this stream...

```
DESCRIBE purchases_enriched;
```

```
Name
                     : PURCHASES_ENRICHED
Field
ROWTIME
                    BIGINT
                                     (system)
                    VARCHAR(STRING) (system)
ROWKEY
                    INTEGER
ORDER_ID
ORDER_DETAILS_ID | INTEGER
ORDER_DATE
                    BIGINT
PRODUCT_ID
                    INTEGER
PRODUCT_NAME
                    VARCHAR(STRING)
PRODUCT_DESC
                  | VARCHAR(STRING)
PRODUCT_COST
                    DECIMAL
PRODUCT_QTY
                    INTEGER
SUPPLIER_ID
                    INTEGER
SUPPLIER_NAME
                    VARCHAR(STRING)
SUPPLIER_EMAIL
                    VARCHAR(STRING)
SUPPLIER_CITY
                    VARCHAR(STRING)
SUPPLIER_COUNTRY |
                    VARCHAR(STRING)
```

...you'll see that we have also denormalized the purchase_orders, purchase_order_details, suppliers and products streams/tables into a single event stream.

Let's query the purchases_enriched stream from the very beginning

```
SET 'auto.offset.reset'='earliest';
SELECT product_id,
    product_name,
    product_qty
FROM purchases_enriched
EMIT CHANGES;
```

Notice that the query returns the first 30 purchase order lines and then stops; this is because no purchase orders are being created by our orders application. The orders application will raise purchase orders for us when we send it some out of stock events.

+		.+
PRODUCT_ID	PRODUCT_NAME	PRODUCT_QTY
+		+
1	Yogurt - Assorted Pack	100
2	Ostrich - Fan Fillet	100
3	Fish - Halibut, Cold Smoked	100
4	Tomatoes Tear Drop Yellow	100
5	Pasta - Fettuccine, Egg, Fresh	100
6	Plastic Wrap	100
7	Pineapple - Regular	100
8	Quail - Eggs, Fresh	100
9	Pork - Ground	100
10	Lamb Shoulder Boneless Nz	100
11	Sausage - Meat	100
12	Herb Du Provence - Primerba	100
13	Bread - Kimel Stick Poly	100
14	Food Colouring - Red	100
15	Cheese - Grie Des Champ	100
16	Longos - Lasagna Veg	100
17	Beets - Golden	100
18	Bread - Dark Rye	100
19	Pepperoni Slices	100
20	Glass - Wine, Plastic, Clear 5 Oz	100
21	Soup - Campbells, Beef Barley	100
22	Bread - Kimel Stick Poly	100
23	Plate - Foam, Bread And Butter	100
24	Parsley - Fresh	100
25	Cookie - Oreo 100x2	100
26	Bread - Crusty Italian Poly	100
27	Wine - Chateauneuf Du Pape	100
28	Country Roll	100
29	Wine - Redchard Merritt	100
30	Doilies - 5, Paper	100



Further Reading

• Stream-Table Joins

Lab 11: Streaming Stock Levels

Before we can create an out of stock event stream, we need to work out the current stock levels for each product. We can do this by combining the sales_enriched stream with the purchases_enriched stream and summing the sales_enriched.quantity column (stock decrements) and the purchases_enriched.quantity column (stock increments).

Let's have a go at this now by creating a new stream called product_supply_and_demand. This stream is consuming messages from the sales_enriched stream and included the product_id and quantity column converted to a negative value, we do this because sales events are our demand and hence decrement stock.

```
SET 'auto.offset.reset'='earliest';
CREATE STREAM product_supply_and_demand WITH (PARTITIONS=1, KAFKA_TOPIC='dc01_product_supply_and_demand') AS SELECT
   product_id,
   product_qty * -1 "QUANTITY"
FROM sales_enriched;
```



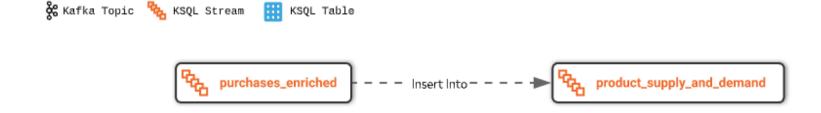
Let's have a quick look at the first few rows of this stream

This query shows a history of all sales and their affect on stock levels.

PRODUCT_ID	QUANTITY	1
+ 1	-6	+
15	-3	1
14	-7	1
23	-3	1
13	-10	1
4	-9	1
10	-9	1
15	-8	1
10	-2	1
27	-7	1
6	-2	
5	-6	
25	-8	
24	-1	
2	-8	1
26	-10	
13	-9	1
16	-9	1
28	-8	1
4	-9	1
Limit Reached		
Query terminated		

What we need to do now is also include all product purchases in the same stream. We can do this using an INSERT INTO statement. The INSERT INTO statement streams the result of a SELECT query into an existing stream and its underlying topic.

```
INSERT INTO product_supply_and_demand
SELECT product_id,
    product_qty "QUANTITY"
FROM purchases_enriched;
```



Our product_supply_and_demand now includes all product sales as stock decrements and all product purchases as stock increments.

We can see the demand for a single product by filtering on the product_id and including only events where the quantity is less than zero.

```
SET 'auto.offset.reset'='earliest';
SELECT product_id,
         quantity
FROM product_supply_and_demand
WHERE product_id = 1
AND quantity < 0
EMIT CHANGES;</pre>
```

PRODUCT_ID	QUANTITY	I
+ 1	-6	+
1	-9	i
1	-7	1
1	-5	1
1	-1	1
1	-7	1
1	-7	
1	-10	1
1	-8	1
1	-4	1
1	-2	1
•••		
•••		
•••		

We can also see the supply for a single product by filtering on the product_id and including only events where the quantity is greater than zero.

This query will only return a single event and reflects the initial purchase order line that was raised for this product.

We're now is a position where we can calculate the current stock level for each product. We can do this by creating a table that groups by the product_id and sums up the quantity column which contains both stock decrements and stock increments.

```
SET 'auto.offset.reset'='earliest';
CREATE TABLE current_stock WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_current_stock') AS SELECT
    product_id
    , SUM(quantity) "STOCK_LEVEL"
```

```
FROM product_supply_and_demand

GROUP BY product_id;

Kafka Topic Kafka Topic KsQL Stream KsQL Table

Froduct_supply_and_demand -- Aggregate Query -- -- current_stock
```

When we guery this table with a Push guery...

...each new event that is displayed on the console reflects the current stock level for the associated product, a new event will be emitted each time a product's stock level changes. Depending on how long it took you to get to this point in the workshop, you may see that all your stock levels are negative. This is because, apart from the initial purchase order for 100 of each product, we have not created any more purchase orders and our customers will have their orders on hold until we acquire more stock, not good, but we'll fix that soon.

 Ω

Further Reading

- INSERT INTO Syntax
- CREATE TABLE AS SELECT Syntax
- KSQL Aggregate Functions

Lab 12: Pull Queries

We can now run our first Pull query. Pull queries are a preview feature in KSQL 5.4 and currently can only be used against tables with aggregates and can only query a single key.

To run a Pull query we just query the table as normal but drop the EMIT CHANGES clause. In this query we are asking "what is the current stock level for product id 1?"

```
select product_id, stock_level from current_stock where rowkey='1';

% Kafka Topic KsQL Stream KsQL Table

Point in Time
Lookups

current_stock
```

The query will return the current stock level and immediatly terminate.

We can also use the KSQL Server's REST endpoint to make Pull gueries.

Exit from the KSQL CLI and run the following from the command line.

```
curl -s -X "POST" "http://localhost:8088/query" -H "Content-Type: application/vnd.ksql.v1+json; charset=utf-8" -d $'{
   "ksql": "select product_id, stock_level from current_stock where rowkey=\'1\';" }'| jq .
```

As you can see, the KSQL Server's REST endpoint will return a JSON message with the product_id and its current stock_level.

This is useful for applications that want access to the current state of the world using a request/response type pattern.



Further Reading

- Pull Queries
- KSQL REST API

Lab 13: Streaming Recent Product Demand

Now that we know the current stock level is for each product, we can use this information to send an event to the orders application and ask it to create purchase orders to replenish the stock, but how much should we stock should we order? we could just order enough to satisfy the current backlog but we'd quickly run out of stock again.

What we really want to do is order enough to satisfy the backlog *and* enough to meet future demand, we can make an attempt at predicting what the future demand will be by looking at the past.

In the following query we are creating a table that will calculate the demand for each product over the last 3 minutes using a window hopping clause.

Hopping windows are based on time intervals. They model fixed-sized, possibly overlapping windows. A hopping window is defined by two properties: the window's duration and its advance, or "hop", interval. The advance interval specifies how far a window moves forward in time relative to the previous window. In our query we we have a window with a duration of three minutes and an advance interval of one minute. Because hopping windows can overlap, a record can belong to more than one such window.

```
SET 'auto.offset.reset'='earliest';
CREATE TABLE product_demand_last_3mins_tbl WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_product_demand_last_3mins')
AS SELECT
    timestamptostring(windowStart(),'HH:mm:ss') "WINDOW_START_TIME"
    , timestamptostring(windowEnd(),'HH:mm:ss') "WINDOW_END_TIME"
    , product_id
```

```
, SUM(product_qty) "DEMAND_LAST_3MINS"
FROM sales_enriched
WINDOW HOPPING (SIZE 3 MINUTES, ADVANCE BY 1 MINUTE)
GROUP BY product_id EMIT CHANGES;
```



If we query this table for a single product...

...you'll see the start and end times for each three minute window, along with the product demand for those 3 minutes. Notice how the window start times are staggered by one minute, this is the advance interval in action. As new sales events occur a new message will be displayed with an update to the window(s) total.

WINDOW_START_	TIME WINDOW_END_TIME	PRODUCT_ID	DEMAND_LAST	Γ_3MINS
+	+	-+	+	+
13:33:00	13:36:00	1	10	
13:34:00	13:37:00	1	10	- 1
13:35:00	13:38:00	1	1	
13:33:00	13:36:00	1	11	- 1
13:34:00	13:37:00	1	11	- 1
13:35:00	13:38:00	1	2	1
13:34:00	13:37:00	1	21	- 1
13:35:00	13:38:00	1	12	- 1
13:36:00	13:39:00	1	10	1
13:34:00	13:37:00	1	26	- 1
13:35:00	13:38:00	1	17	- 1
13:36:00	13:39:00	1	15	- 1
13:35:00	13:38:00	1	22	
13:36:00	13:39:00	1	20	- 1
13:37:00	13:40:00	1	5	- 1
13:36:00	13:39:00	1	28	- 1
13:37:00	13:40:00	1	13	
13:38:00	13:41:00	1	8	1

We will now create a stream from this table and then join it to the current_stock table

Create a stream from the table's underlying topic...

```
CREATE STREAM product_demand_last_3mins WITH (KAFKA_TOPIC='dc01_product_demand_last_3mins', VALUE_FORMAT='AVRO');
```





Further Reading

• Windows in KSQL Queries

Lab 14: Streaming "Out of Stock" Events

Now that we have the current_stock table and product_demand_last_3mins stream, we can create a out_of_stock_events stream by joining the two together and calculating the required purchase order quantity. We calculate the purchase_qty from adding the inverse of the current stock level to the last 3 minutes of demand. The stream is filtered to only include products that have no stock and therefore need purchase orders raising for them.

```
SET 'auto.offset.reset' = 'latest';

CREATE STREAM out_of_stock_events WITH (PARTITIONS = 1, KAFKA_TOPIC = 'dc01_out_of_stock_events')

AS SELECT

cs.product_id "PRODUCT_ID",

pd.window_start_time,

pd.window_end_time,

cs.stock_level,

pd.demand_last_3mins,

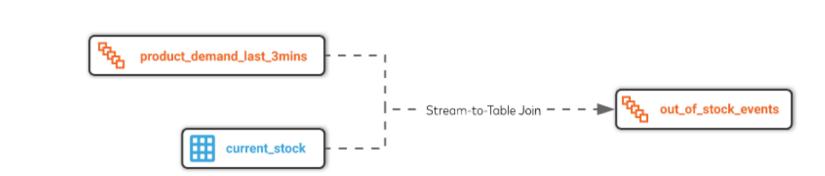
(cs.stock_level * -1) + pd.DEMAND_LAST_3MINS "QUANTITY_TO_PURCHASE"

FROM product_demand_last_3mins pd

INNER JOIN current_stock cs ON pd.product_id = cs.product_id

WHERE stock_level <= 0;

KSQL Stream KSQL Table
```



When we query the out_of_stock_events stream...

```
SET 'auto.offset.reset' = 'latest';
SELECT product_id,
    window_start_time,
    window_end_time,
    stock_level,
    demand_last_3mins,
    quantity_to_purchase
FROM out_of_stock_events
EMIT CHANGES;
```

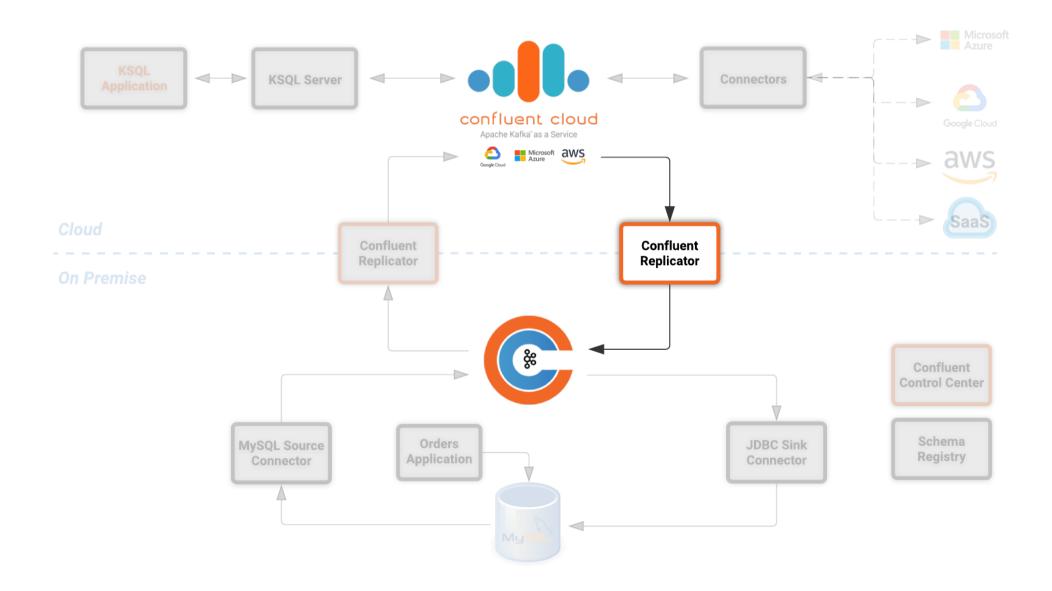
...you'll see a constant stream of *out of stock products* and the predicted purchase quantity that should be ordered to satisfy any current backlog and also meet the next 3 minutes demand.

```
+-----+
|PRODUCT_ID |WINDOW_START_TIME |WINDOW_END_TIME |STOCK_LEVEL |DEMAND_LAST_3MINS|QUANTITY_TO_PURCASE|
```

28	13:53:00	13:56:00	-85	12	97	
28	13:54:00	13:57:00	-85	1	86	1
28	13:55:00	13:58:00	-85	1	86	
4	13:53:00	13:56:00	-128	26	154	1
4	13:54:00	13:57:00	-128	11	139	
4	13:55:00	13:58:00	-128	11	139	
5	13:53:00	13:56:00	-73	15	88	
5	13:54:00	13:57:00	-73	15	88	1
5	13:55:00	13:58:00	-73	15	88	1
28	13:53:00	13:56:00	-85	18	103	1
28	13:54:00	13:57:00	-91	7	98	1
28	13:55:00	13:58:00	-91	7	98	1
14	13:53:00	13:56:00	-156	31	187	1
14	13:54:00	13:57:00	-156	15	171	1
14	13:55:00	13:58:00	-156	6	162	1
5	13:53:00	13:56:00	-73	25	98	
5	13:54:00	13:57:00	-83	25	108	
5	13:55:00	13:58:00	-83	25	108	
12	13:53:00	13:56:00	-197	25	222	
12	13:54:00	13:57:00	-197	21	218	
		13:58:00	-200	3	203	

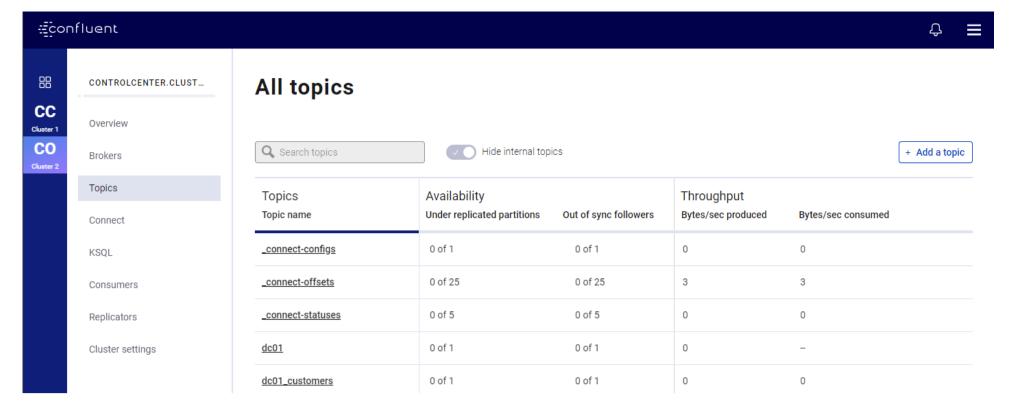
Lab 15: Replicate Events to On-Premise Kafka

The next step is to push the out_of_stock_events stream to our application so it can create some purchase orders for us. To do this we'll need to replicate the dc01_out_of_stock_events topic from Confluent Cloud back to our on-premise Kafka cluster.

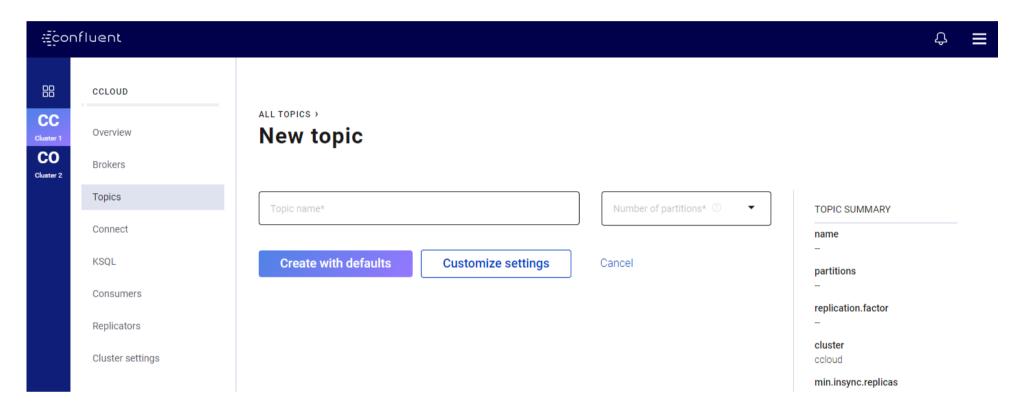


Before we do that, let's create the target topic in our on-premise Kafka cluster using Confluent Control Center

Select your on-premise cluster from the left-hand navigation bar, select "topics" and then click on "Add a Topic".



Name the topic dc01_out_of_stock_events and click "Create with defaults"



We are now ready to replicate this topic from Confluent Cloud to you on-premise cluster.

Submit the Replicator Connector Config

Execute the following from the command line to create the Replicator Connector. You can see that we have asked to only replicate the dc01_out_of_stock_events topic by configuring "topic.whitelist": "dc01_out_of_stock_events"

```
curl -i -X POST -H "Accept:application/json" \
    -H "Content-Type:application/json" http://localhost:18083/connectors/ \
    -d '{
        "name": "replicator-ccloud-to-dc01",
        "config": {
          "connector.class": "io.confluent.connect.replicator.ReplicatorSourceConnector",
          "key.converter": "io.confluent.connect.replicator.util.ByteArrayConverter",
          "value.converter": "io.confluent.connect.replicator.util.ByteArrayConverter",
          "topic.config.sync": "false",
          "topic.whitelist": "dc01_out_of_stock_events",
          "dest.kafka.bootstrap.servers": "broker:29092",
          "dest.kafka.replication.factor": 1,
          "src.kafka.bootstrap.servers": "${file:/secrets.properties:CCLOUD_CLUSTER_ENDPOINT}",
          "src.kafka.security.protocol": "SASL_SSL",
          "src.kafka.sasl.mechanism": "PLAIN",
          "src.kafka.sasl.jaas.config": "org.apache.kafka.common.security.plain.PlainLoginModule required
username=\"${file:/secrets.properties:CCLOUD_API_KEY}\" password=\"${file:/secrets.properties:CCLOUD_API_SECRET}\";",
          "src.consumer.group.id": "replicator-ccloud-to-dc01",
          "src.consumer.interceptor.classes":
"io.confluent.monitoring.clients.interceptor.MonitoringConsumerInterceptor",
```

You should see something similar...

```
HTTP/1.1 100 Continue

HTTP/1.1 201 Created

Date: Sun, 09 Feb 2020 15:07:22 GMT

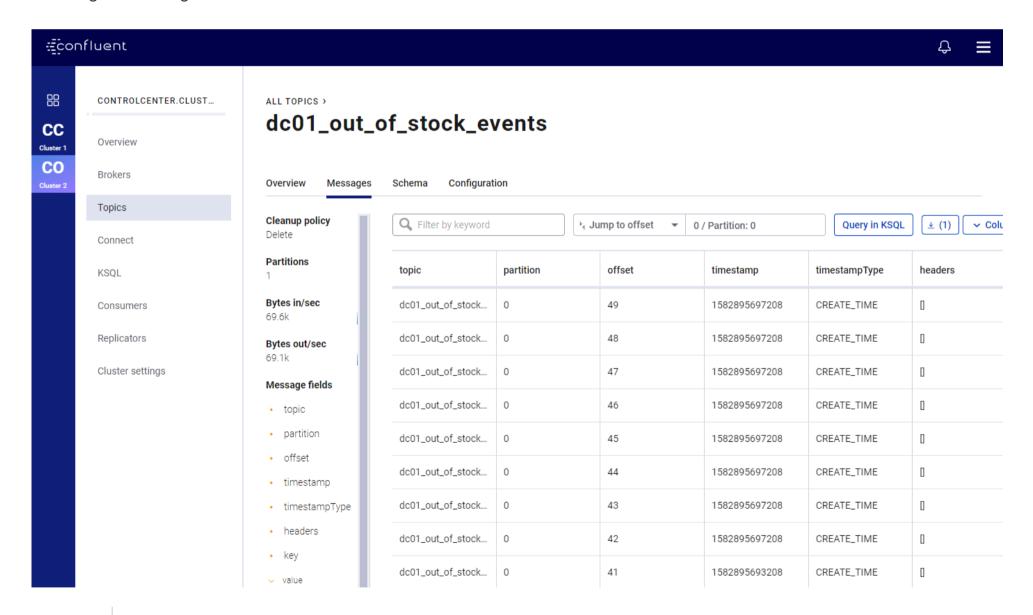
Location: http://localhost:18084/connectors/replicator-dc01-to-ccloud

Content-Type: application/json

Content-Length: 1342

Server: Jetty(9.4.20.v20190813)
...
...
```

We can confirm that the dc01_out_of_stock_events is being replicated from Confluent Cloud to our on-premise cluster by checking for messages in Confluent Control Center



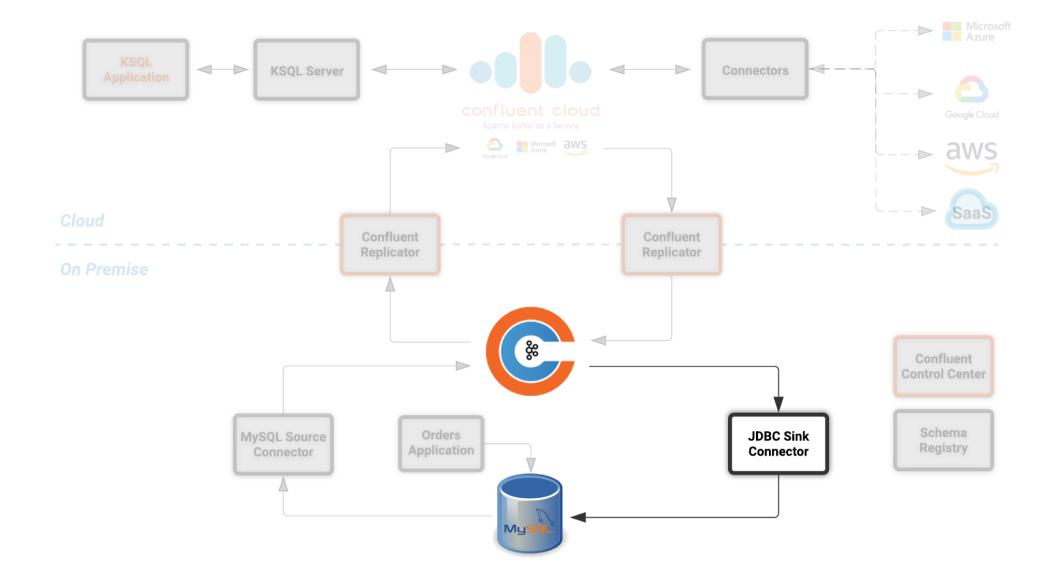


Further Reading

- Confluent Replicator
- Confluent Replicator Configuration Properties

Lab 16: Sink Events into MySQL

Finally we need to sink the dc01_out_of_stock_events topic into a MySQL database table, the on-premise application will then process these events and create purchase order for us.



But before we do that, let's open a couple more terminal sessions and start the KSQL CLI in each.

```
ssh dc01@34.89.105.9

docker exec -it ksql-cli ksql http://ksql-server-ccloud:8088
```

Execute the following query in the 1st session...

...and this query in the 2nd session

```
SET 'auto.offset.reset'='latest';
SELECT product_id,
     product_qty
FROM purchases_enriched
EMIT CHANGES;
```

You now have a real time view of the current product stock levels in the first KSQL session and the purchases being made to replenish the stock in second. Not that the second query isn't returning anything yet.

Let's now sink the *out of stock events* to the MySQL database using the JDBC Connector. Once the events start arriving in the database, the orders application will process them and start generating the required purchase orders.

In a third terminal session, create the JDBC Sink Connector by running the following from the command line.

```
curl -i -X POST -H "Accept:application/json" \
   -H "Content-Type:application/json" http://localhost:18083/connectors/ \
```

```
-d '{
    "name": "jdbc-mysql-sink",
    "config": {
        "connector.class": "io.confluent.connect.jdbc.JdbcSinkConnector",
        "topics": "dc01_out_of_stock_events",
        "connection.url": "jdbc:mysql://mysql:3306/orders",
        "connection.user": "mysqluser",
        "connection.password": "mysqlpw",
        "insert.mode": "INSERT",
        "batch.size": "3000",
        "auto.create": "true",
        "key.converter": "org.apache.kafka.connect.storage.StringConverter"
    }
}'
```

Observe the current stock query in the first KSQL session, when a product has zero or less stock you should see a purchase event appear in the second KSQL session and then the new stock level reflected in the first session. In theory, given a constant demand, each product should run out of stock and get replenished roughly every 3 minutes.



Further Reading

- JDBC Sink Connector
- JDBC Sink Connector Configuration Properties

Optional Lab: Stream Sales & Purchases to Google Cloud Storage

We can use the Google Cloud Storage Sink Connector to stream changes from a topics to Google Cloud Storage, from here the data can be consumed other Google Cloud Platform services.

Another preview feature of KSQL 5.4 is the ability to create connectors directly within KSQL.

Start a KSQL CLI session

```
docker exec -it ksql-cli ksql http://ksql-server-ccloud:8088
```

And run the following CREATE SINK CONNECTOR command. This will create a connector that will sink the dc01_sales_enriched and the dc01_purchases_enriched topics to Google Cloud Storage.

```
CREATE SINK CONNECTOR dc01_gcs_sink WITH (
  'connector.class' = 'io.confluent.connect.gcs.GcsSinkConnector',
  'tasks.max'= '1',
  'gcs.bucket.name' = '${file:/secrets.properties:GCS_BUCKET_NAME}',
  'gcs.part.size' = '5242880',
  'gcs.credentials.path' = '${file:/secrets.properties:GCS CREDENTIALS PATH}',
  'storage.class'= 'io.confluent.connect.gcs.storage.GcsStorage',
  'format.class'= 'io.confluent.connect.gcs.format.avro.AvroFormat',
  'partitioner.class'= 'io.confluent.connect.storage.partitioner.DefaultPartitioner',
  'schema.compatibility'= 'NONE',
  'topics' = 'dc01_sales_enriched,dc01_purchases_enriched',
  'confluent.topic.bootstrap.servers'= '${file:/secrets.properties:CCLOUD_CLUSTER_ENDPOINT}',
  'confluent.topic.security.protocol' = 'SASL_SSL',
  'confluent.topic.sasl.mechanism' = 'PLAIN',
  confluent.topic.sasl.jaas.config' = 'org.apache.kafka.common.security.plain.PlainLoginModule required'
username=\"${file:/secrets.properties:CCLOUD API KEY}\" password=\"${file:/secrets.properties:CCLOUD API SECRET}\";',
  'confluent.topic.replication.factor'= '3',
  'producer.interceptor.classes' = 'io.confluent.monitoring.clients.interceptor.MonitoringProducerInterceptor',
  'key.converter'= 'org.apache.kafka.connect.storage.StringConverter'
);
```

We can list our current connectors using the following command

show connectors;

```
Connector Name | Type | Class

DC01_GCS_SINK | SINK | io.confluent.connect.gcs.GcsSinkConnector

replicator-dc01-to-ccloud | SOURCE | io.confluent.connect.replicator.ReplicatorSourceConnector
```

We can also describe a connector and view its status using the describe connector statement.

describe connector DC01_GCS_SINK;

Name : DC01_GCS_SINK

Class : io.confluent.connect.gcs.GcsSinkConnector

Type : sink
State : RUNNING

WorkerId : kafka-connect:18084

Task ID | State | Error Trace

O | RUNNING |

Depending on who's hosting the workshop, you may or may not have access to the GCP account where the storage bucket is held.



Further Reading

- Google Cloud Storage Sink Connector
- Google Cloud Storage Sink Connector Configuration Properties

Optional Lab: Stream Sales & Purchases to Google Big Query

We can use the BigQuery Sink Connector to stream changes from a topics to Google BigQuery.

Another preview feature of KSQL 5.4 is the ability to create connectors directly within KSQL.

Start a KSQL CLI session

```
docker exec -it ksql-cli ksql http://ksql-server-ccloud:8088
```

And run the following CREATE SINK CONNECTOR command. This will create a connector that will stream the following topics to Google BigQuery:-

```
dc01_sales_orders
dc01_sales_order_details
dc01_purchase_orders
dc01_purchase_order_details
dc01_products
dc01_customers
dc01_suppliers
```

```
CREATE SINK CONNECTOR dc01_gbq_sink WITH (
   'connector.class'='com.wepay.kafka.connect.bigquery.BigQuerySinkConnector',
   'schemaRetriever'='com.wepay.kafka.connect.bigquery.schemaregistry.schemaretriever.SchemaRegistrySchemaRetriever',
   'schemaRegistryLocation'= 'http://schema-registry:8081',
   'topics'=
'dc01_sales_orders,dc01_sales_order_details,dc01_purchase_orders,dc01_purchase_order_details,dc01_products,dc01_custo
mers,dc01_suppliers',
   'tasks.max'='1',
   'sanitizeTopics'='true',
   'autoCreateTables'='true',
   'autoUpdateSchemas'='true',
   'project'='${file:/secrets.properties:GBQ_PROJECT}',
   'datasets'='.*=${file:/secrets.properties:GBQ_DATASET}',
   'keyfile'='${file:/secrets.properties:GBQ_CREDENTIALS_PATH}'
);
```

We can list our current connectors using the following command

```
Connector Name | Type | Class

DC01_GBQ_SINK | SINK | com.wepay.kafka.connect.bigquery.BigQuerySinkConnector
replicator-dc01-to-ccloud | SOURCE | io.confluent.connect.replicator.ReplicatorSourceConnector
```

We can also describe a connector and view its status using the describe connector statement.

```
describe connector DC01_GBQ_SINK;

Name : DC01_GBQ_SINK
```

Depending on who's hosting the workshop, you may or may not have access to the GCP account where the BigQuery dataset is held.

Visualize your Data in Google Data Studio

Now that your Data is in BigQuery, you can use Google Datastudio to visualize it.

Open Goodle Data Studio and create a new Report. Add new Datasources and select BigQuery. You can use the queries below for your convenience (look for the Custom Query in the left sidebar).

In the gueries replace the following according to your environment:

- gcp-project-id: your GCP project ID, where BigQuery stores the data
- bigquery_dataset: Your Big Query dataset name

```
SELECT
   SUM(OD.quantity) as order_quantity,
   AVG(P.price) as avg_product_price,
   P.name as product_name
FROM `gcp-project-id.bigquery_dataset.dc01_sales_order_details` OD
INNER JOIN `gcp-project-id.bigquery_dataset.dc01_products` P ON OD.product_id = P.id and P.sourcedc="dc01"
WHERE
OD.sourcedc="dc01"
GROUP BY P.name
```

Top customers

```
SELECT

COUNT(DISTINCT 0.id) as order_count,
SUM(OD.quantity) as order_quantity,
SUM(OD.price) as order_price,
C.first_name || " " || C.last_name as customer_name
FROM `gcp-project-id.bigquery_dataset.dc01_sales_orders` O
INNER JOIN `gcp-project-id.bigquery_dataset.dc01_sales_order_details` OD ON 0.id = OD.sales_order_id and
OD.sourcedc="dc01"
INNER JOIN `gcp-project-id.bigquery_dataset.dc01_customers` C ON 0.customer_id = C.id and C.sourcedc="dc01"
WHERE
O.sourcedc="dc01"
GROUP BY customer_name
```

Top suppliers

```
SELECT

COUNT(DISTINCT 0.id) as order_count,
SUM(OD.quantity) as order_quantity,
SUM(OD.price) as order_price,
S.name AS supplier_name

FROM `gcp-project-id.bigquery_dataset.dc01_sales_orders` O

INNER JOIN `gcp-project-id.bigquery_dataset.dc01_sales_order_details` OD ON 0.id = OD.sales_order_id and
OD.sourcedc="dc01"

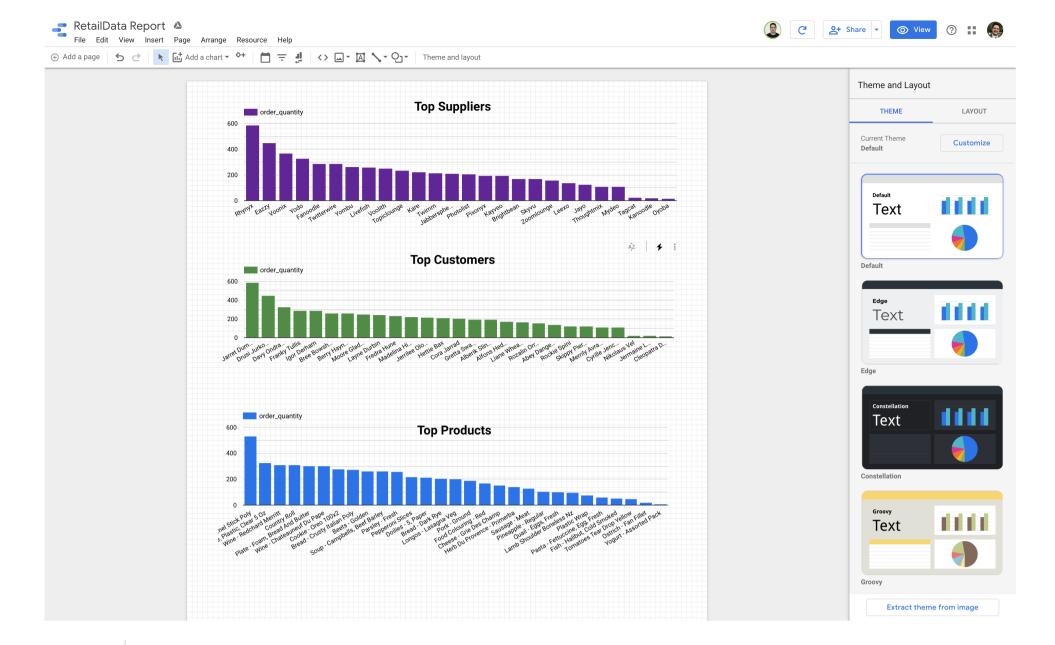
INNER JOIN `gcp-project-id.bigquery_dataset.dc01_suppliers` S ON 0.customer_id = S.id and S.sourcedc="dc01"

WHERE

0.sourcedc="dc01"

GROUP BY supplier_name
```

This is an example of a report you can build:



$\mathbf{\Omega}$

Further Reading

- Google BigQuery Sink Connector
- Google BigQuery Sink Connector Configuration Properties

Optional Lab: Stream Sales & Purchases to MongoDB Atlas

We can use the MongoDB Sink Connector to stream changes from Confluent Cloud to MongoDB Atlas, from here the data can be leveraged in the wider MongoDB ecosystem.

Another preview feature of KSQL 5.4 is the ability to create connectors directly within KSQL.

Start a KSQL CLI session

```
docker exec -it ksql-cli ksql http://ksql-server-ccloud:8088
```

And run the following CREATE SINK CONNECTOR command. This will create a connector that will sink the dc01_sales_enriched and the dc01_purchases_enriched topics to MongoDB.

```
CREATE SINK CONNECTOR dc01_mongodb_sink WITH (
    'connector.class'='com.mongodb.kafka.connect.MongoSinkConnector',
    'tasks.max'='1',
    'topics'='dc01_sales_enriched,dc01_purchases_enriched',
    'connection.uri'='${file:/secrets.properties:MONGODBATLAS_SRV_ADDRESS}',
    'database'='demo',
    'collection'='dc01',
    'topic.override.dc01_sales_enriched.collection'='dc01_sales',
    'topic.override.dc01_purchases_enriched.collection'='dc01_purchases',
    'key.converter'='org.apache.kafka.connect.storage.StringConverter',
    'transforms'='WrapKey',
    'transforms.WrapKey.type'='org.apache.kafka.connect.transforms.HoistField$Key',
    'transforms.WrapKey.field'='ROWKEY',
    'document.id.strategy'='com.mongodb.kafka.connect.sink.processor.id.strategy.UuidStrategy',
```

```
'post.processor.chain'='com.mongodb.kafka.connect.sink.processor.DocumentIdAdder',
   'max.batch.size'='20'
);
```

We can list our current connectors using the following command

show connectors;

```
Connector Name | Type | Class

dc01_MONGODB_SINK | SINK | com.mongodb.kafka.connect.MongoSinkConnector

replicator-dc01-to-ccloud | SOURCE | io.confluent.connect.replicator.ReplicatorSourceConnector
```

We can also describe a connector and view its status using the describe connector statement.

describe connector dc01_MONGODB_SINK;

Name : dc01_MONGODB_SINK

Class : com.mongodb.kafka.connect.MongoSinkConnector

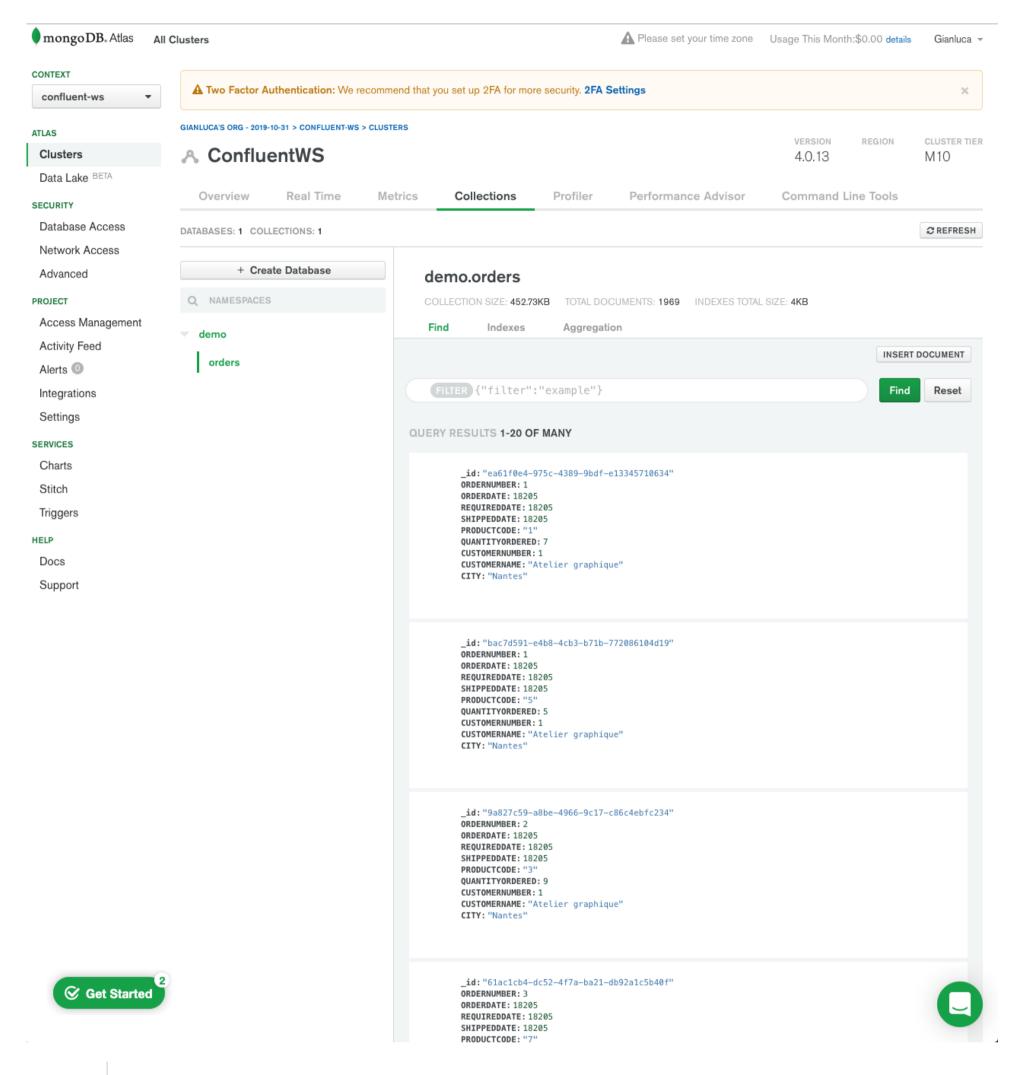
Type : sink
State : RUNNING

WorkerId : kafka-connect:18084

Task ID | State | Error Trace

0 | RUNNING |

Depending on who's hosting the workshop, you may or may not have access to the MongoDB Atlas account where the database is held.





Further Reading

- MongoDB Atlas
- MongoDB Kafka Connector

Wrapping up

During this workshop we have seen how the Confluent Platform and Confluent Cloud can be used to build event driven, real time applications that span the data center and public cloud.

Last updated 2020-03-19 20:48:17 UTC