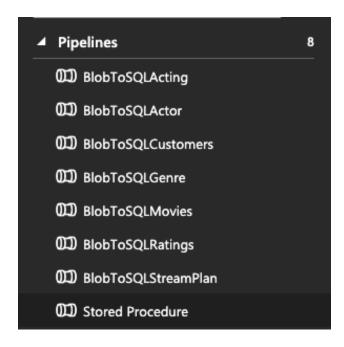
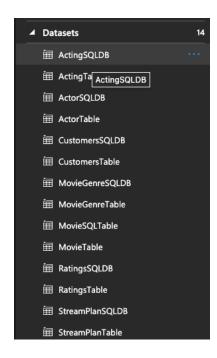
P4- Implementation Streaming Service Analysis Team 6

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In this phase of our project implementation, we have utilized linked services in Azure Data Factory to connect and manage data across different storage and database services. As we now have cleaned datasets that we transformed using Alteryx and Python in Phase-3, we established linked services to Azure Blob Storage and Azure SQL Database, which acted as the sources and destinations for our data. Blob Storage was primarily used for handling large volumes of structured + unstructured data, whereas SQL Database served as our structured data's destination storage system.

The core of our Azure Data Factory implementation involved setting up Data Pipelines and Scheduled Triggers. These pipelines were designed to efficiently transfer and transform data between the source and sink database, "xxxTable" & "xxxSQLDB", respectively.





Showing 1 - 11 items				
Pipeline name	Run start ↑	Run end	Duration	Status 1
	12/7/2023, 9:48:47 PM	12/7/2023, 9:48:51 PM	5s	Succeeded
	12/7/2023, 9:41:27 PM	12/7/2023, 9:41:44 PM	18s	✓ Succeeded
	12/7/2023, 7:19:52 PM	12/7/2023, 7:20:08 PM	17s	✓ Succeeded
	12/7/2023, 7:07:25 PM	12/7/2023, 7:07:42 PM	17s	Succeeded
	12/7/2023, 4:20:07 PM	12/7/2023, 4:20:24 PM	18s	✓ Succeeded
	12/7/2023, 3:49:27 PM	12/7/2023, 3:49:44 PM	18s	⊘ Succeeded
	12/7/2023, 3:45:12 PM	12/7/2023, 3:45:29 PM	18s	✓ Succeeded
	12/7/2023, 3:38:33 PM	12/7/2023, 3:38:51 PM	18s	✓ Succeeded
	12/7/2023, 3:34:01 PM	12/7/2023, 3:34:18 PM	18s	✓ Succeeded
	12/7/2023, 12:33:51 AM	12/7/2023, 12:34:11 AM	21s	Succeeded
	12/7/2023, 12:03:31 AM	12/7/2023, 12:04:19 AM	48s	⊘ Succeeded

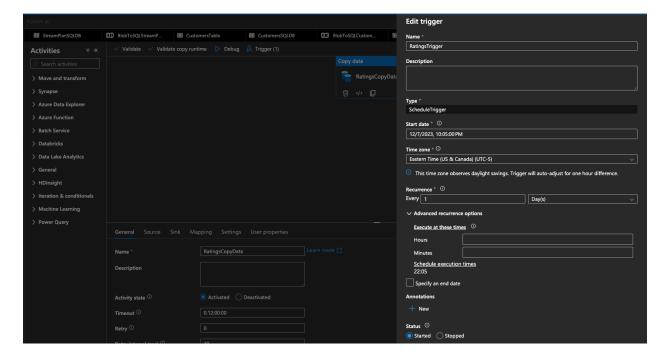
We focused on creating a seamless flow for specific datasets, particularly our document model (Customers, Stream-Plan & Movie-Ratings) as it. This document represents each Customer's info, the stream plan they are on and the Movies they have rated on IMDB. To do the same, we wrote a Stored Procedure (code below) and scheduled an **Ongoing Data Refresh** that executes this SP on a 24-hour basis. This was also done for our Customers table.

```
CREATE PROCEDURE GetCustomerDetails
BEGIN
    SELECT
        c.CustID as id,
        c.Cust_FName as firstName,
        c.Cust_LName as lastName,
        c.Email as email,
        sp.Description as streamPlanDescription,
        sp.Monthly_Cost as streamPlanMonthlyCost,
        (SELECT
             r.MovieID as movieID,
             r.IMDB_Rating as rating
         FROM Ratings r
         WHERE r.CustID = c.CustID
         FOR JSON PATH) as ratedMovies
    FROM Customers c
    INNER JOIN Stream_Plan sp ON sp.Stream_Plan_ID = c.Stream_Plan_ID
    FOR JSON PATH;
END;
```

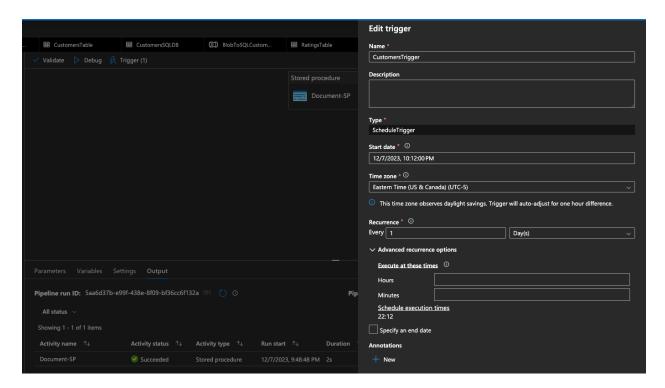
Generated JSON Output (Customers, Stream-Plan & Movie-Ratings):

To manage and automate the entire data flow, we set up triggers in ADF. These triggers were configured to initiate pipeline runs based on our requirements, ensuring that data processing was timely and consistent. This setup not only streamlined our data management tasks but also provided a scalable and flexible solution for handling our data processing needs in ADF.

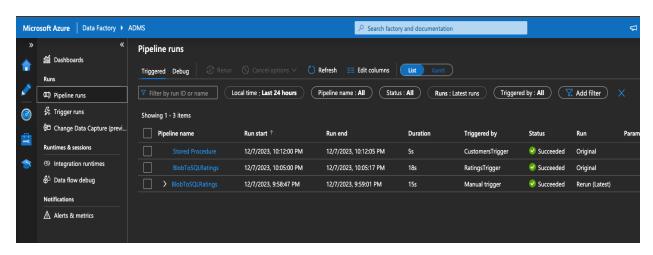
Ratings Table will need updating regularly as many people are more likely to rate many movies, so this trigger updates the same every 24 hours.



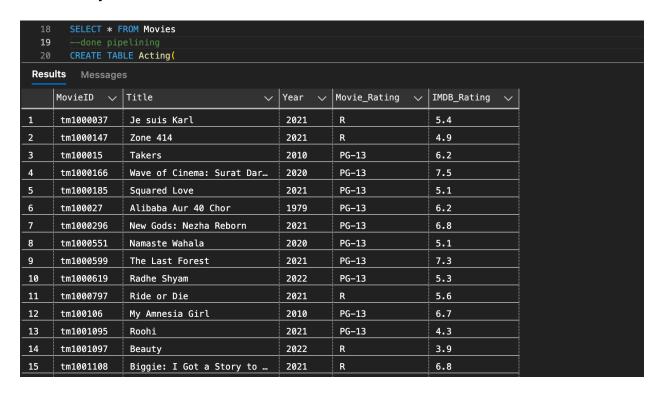
Similarly, for Customers table as new customers can join the platform every day:



Successful runs of our Scheduled and Manual Triggers:



Here are some reference screenshots to show that the tables update & populated successfully:



Row Count validation:

