## Analytics on Streaming Data with Azure Stream Analytics

Data Science Dojo





## Introducing Big Data

Continued

Exabytes (10E18)

Petabytes (10E15)

Terabytes (10E12)

Gigabytes (10E9)



Velocity - Variety

ERP / CRM

WEB

Internet of things

## **Defining Real-time**

Within seconds...

or...

Within minutes...

of an event occurring

Up to 2 hours



#### **Timeliness of Information**



What was trending in the past 5 minutes?

Amber alert car detected!





A tornado will form in the next 30 minutes.



#### **Timeliness of Information**



A stock is going to crash in 20 minutes.

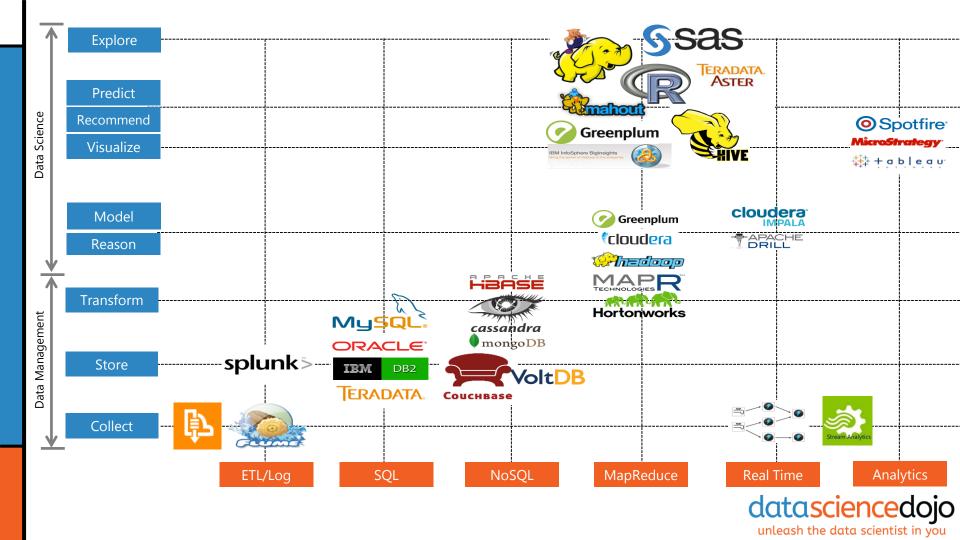
A fire is about to start in your house.





The power grid will overload in 2 minutes.





## **Typical Event Processing**





**Applications** Cloud Gateways (WebAPIs)





Scalable **Event Broker** 

**Event Hub** 



Real-Time Analytics



External **Data Sources** 



Web/Thick Client Dashboards



Search And Query







(Extract, Transform, Load)



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## **Typical Event Processing**





Q

**Devices** 





Cloud Gateways (WebAPIs)



Scalable Event Broker



Real-time Analytics



External Data Sources



Web/Thick Client Dashboards



Search And Query







**Stream Analytics** 



**Data Analytics** 



#### When to use Stream Processors





#### ETL Should Still Happen



**Stream Processing** 

#### ETL

- Stream Processing is only icing on the cake.
- It can, but should not replace a company's normal ETL cycle.



#### Popular Up and Coming Event Processors





**Google DataFlow** 









#### Demo



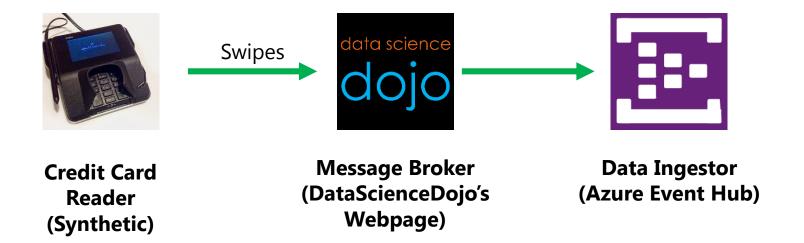
#### **Credit Card Transactions (swipes)**



- Credit card transactions are usually done in batch as an EOTD send.
- Stream process for insights now.
- US mainland transactions



## Previously...





#### The Streamer

http://dojodemos.azurewebsites.net/credit-card-streamer/

#### Credit Card Streamer

This app will simulate the kind of data streams that banks would encounter, credit card swipe data. The app will generate synthetic data from a credit card transaction (swipe) and pushes it into a given Azure Event Hub as a JSON. The application logic for this app is written entirely in JavaScript so the speed and interval of the transactions is dependent on the processing power of the user device.

△ Event Hub Credentials		
Event Hub Name (Need help? PDF Guide)		
field required		
Service Bus Namespace (Need help? PDF Guide)		
field required		
Shared Access Policy Name (Need help? PDF Guide)		
field required		

■ Output Preview	
Display Format (Data is still sent as a JSON):	JSON ⟨/> List !!!
Successfully loaded database.	Ready to simulate data.



#### The Data

```
"swipe_date":"2015-05-22T20:16:27.122Z",
"transaction id":3127484,
"card type":"VISA",
"card number":"4913419738164560",
"expiration_month":"02",
"expiration year":"18",
"cvv code":"520",
"user id":"972288",
"user gender": "male",
"user first_name":"Alexander",
"user last name":"Hamilton",
"merchant": "McDonald's",
"transaction_amount":13.64,
"balance":336.48,
"merchant fee":.5,
"swipe city":"New York",
"swipe state":"New York",
"swip_city_state":"New York, NY",
"InstanceNo":1
```

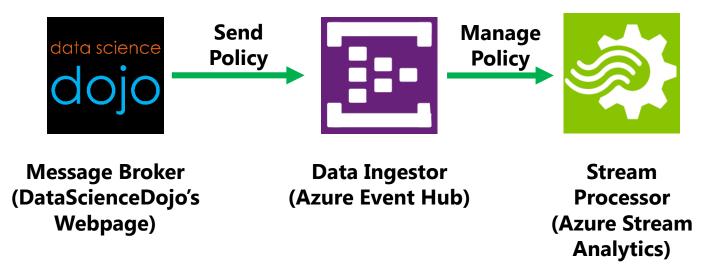
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#### Inside the Event Hub



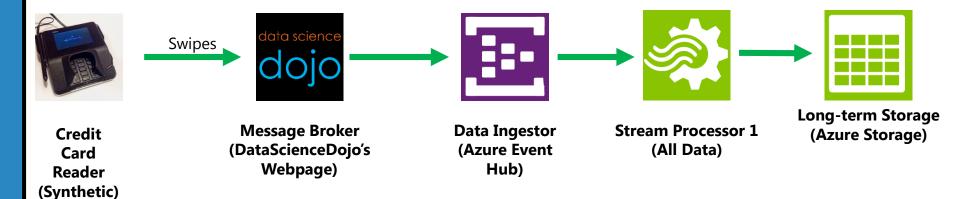


## **Setting Policies**





#### With Stream Processor



#### **More Processors Dashboard** (PowerBI) Swipes **Long-term Storage Message Broker Data Ingestor Stream Processor 1** Credit (Azure Storage) (DataScienceDojo's (Azure Event (All Data) Card Webpage) Hub) Reader (Synthetic) **Stream Processor 2 Anomaly Detection** (Filter/Aggregate) **Event Hub**

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#### **SQL** with Data at Rest

- Question "Show me VISA transactions from last month."
- Answering with a relational database No problem! Here you go!
- SELECT \*
   FROM credit\_db
   WHERE card\_type like VISA'



#### **SQL** Data in Motion

- Different Question "Show me VISA transactions in the past 2 minutes."
- Answering with a relational database
   I'm not ready yet... Ask again later.... Or tomorrow (after batch)...
- Not a great solution...





#### Azure Stream Query Language

- Queries through time
- Simple SQL dialect
  - Familiar learning curve reduction
  - High-Level expression of intent, not implementation
  - Maintainable focus on the essentials of the problem
- Extended in natural ways to express temporal concepts
  - WINDOW multiple kinds
    - Tumbling, hopping, sliding
  - TIMESTAMP BY, BETWEEN
  - DATEDIFF in joins
  - PARTITION BY for scale-out

```
WITH agg AS
(
     SELECT Avg(reading), Building
     FROM Temperature
     GROUP BY TumblingWindow(minute, 1), building
)
SELECT A1.Avg AS Old, A2.Avg AS New, A1.Building
FROM Agg A1 JOIN Agg A2
ON A1.Building = A2.Building
AND DATEDIFF(minute,A1,A2) BETWEEN 4.5 AND 5.5
WHERE
     (a1.avg < a2.avg - 10) OR (a1.avg > a2.avg+10)
```



#### **Temporal System**

- Every event is a point in time, and thus must come with a timestamp
  - Remember how relational DBs need a PK? Temporal systems need a timestamp as its unique identifier.
  - Temporal integrity and referential integrity
- Stream Analytics can append your events with a timestamp (bad practice if standalone)
  - The default timestamp will be when the event enters Stream Analytics
  - Can be skewed by network and hardware latency, or legacy processing
- Users can define application time stamps with the TIMESTAMP BY clause



#### Which Timestamp?

```
"swipe_date": "2015-05-21T22:47:55.0770000Z", <
"transaction id":222301082.
"card_type":"VISA",
"card_number":"40265691066025560",
"expiration_month":"06",
"expiration_year":"22",
"cvv_code":"3310",
"user_id":"690548",
"user_gender":"male",
"user_first_name":"Caden",
"user_last_name":"Hatton",
"merchant":"Macy's",
"transaction amount":4.98,
"balance":7223.9,
"merchant fee":0.5,
"swipe_city":"New York",
"swipe_state":"New York",
"swip_city_state":"New York, NY",
"InstanceNo":1,
"EventProcessedUtcTime": "2015-05-21T22:47:50.0879821Z"
"PartitionId":3,
"EventEnqueuedUtcTime":"2015-05-21T22:47:49.9850000Z
```

Time of event

Time processed by stream processor

Time entered broker



#### Same Event...

```
{
    "swipe_date":"2015-05-21T22:47:55.0770000Z",
    "EventProcessedUtcTime":"2015-05-21T22:47:50.0879821Z",
    "EventEnqueuedUtcTime":"2015-05-21T22:47:49.9850000Z"
}
```

According to these timestamps, the event happened 5 seconds AFTER the event was processed and queued.

- How can that be?
- The event was not confined to the physical laws of space and time.

#### The clock on your device matters.



#### Azure Stream Query Language

Show me transactions as they happen.
 Write it to a blob AND powerBI.

```
INTO MyBlob
FROM SwipeStream TIMESTAMP BY swipe_date;
SELECT *
INTO PowerBI
FROM SwipeStream TIMESTAMP BY swipe_date;
```



#### StreamQL: Calculations

What was our commission on each transaction?

```
SELECT
```

```
transaction_id,
merchant_fee / transaction_amount AS Commision
FROM SwipeStream
TIMESTAMP BY swipe_date
```



#### StreamQL: Filter Queries

Show me only VISA transactions that made over \$5 revenue.

# swipe\_date, card\_type, merchant\_fee AS revenue FROM SwipeStream TIMESTAMP BY swipe\_date WHERE card\_type LIKE 'VISA' AND merchant\_fee < 5

SWIPE_DATE	CARD_TYPE	REVENUE
2015-05-21T2	VISA	6.2
2015-05-21T2	VISA	10.31
2015-05-21T2	VISA	11.72
2015-05-21T2	VISA	7.82
2015-05-21T2	VISA	9.91
2015-05-21T2	VISA	7.62
2015-05-21T2	VISA	5.25



#### **Temporal Questions**

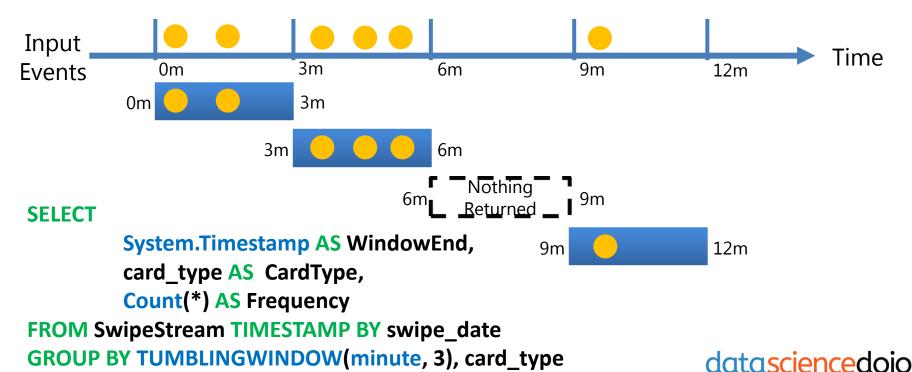
Count the number of transactions....

- When should the counting of transactions begin?
- When should the counting of transactions end?
- How long should the transactions be counted for?
- How often do transactions need to be counted?



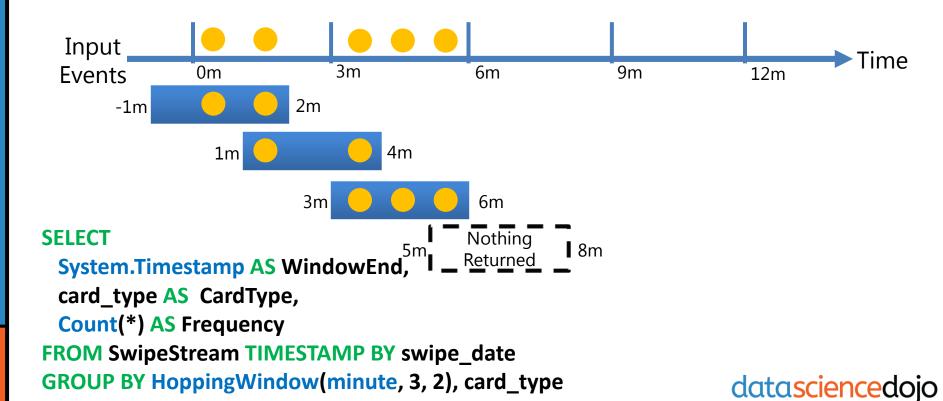
## **Tumbling Window**

How many transactions were made for each card type every minute?



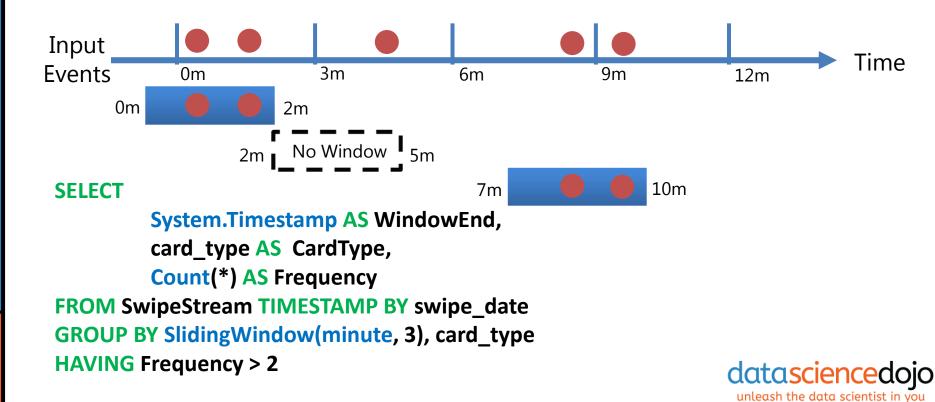
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## **Hopping Window**



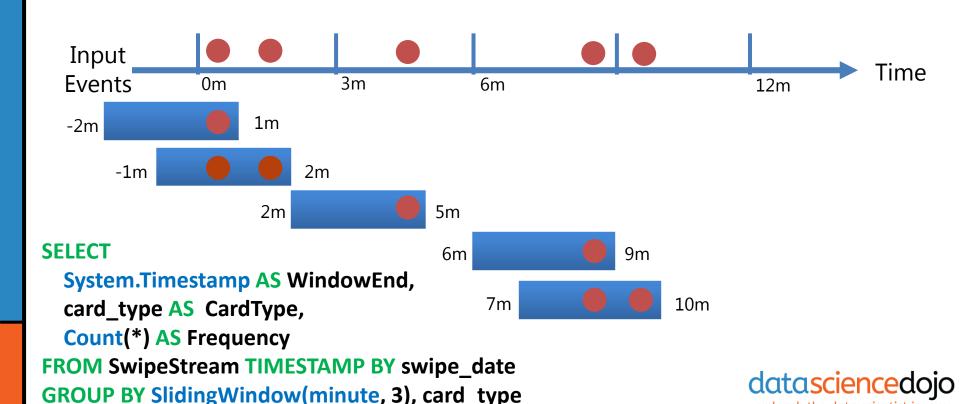
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## Sliding Window



Time

#### Sliding Window: Without 'Having' Clause



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## **Sum Aggregation**

• How much revenue is being accumulated from merchants every 3 minutes?

#### **SELECT**

```
System.Timestamp AS WindowEnd,
Sum(merchant_fee) AS IntervalRevenue
FROM SwipeStream TIMESTAMP BY swipe_date
GROUP BY TUMBLINGWINDOW(minute, 3), WindowEnd
```



#### Sum Aggregation: With Filtering

• Which 3-minute time interval made more than \$10?

```
SELECT
```

```
System.Timestamp AS WindowEnd,
Sum(merchant_fee) AS IntervalRevenue
FROM SwipeStream TIMESTAMP BY swipe_date
GROUP BY TUMBLINGWINDOW(minute, 3), WindowEnd
Having IntervalRevenue > 10
```



#### **Descriptive Statistics**

 Generate descriptive statistics for revenue every 3 minutes (car count, min, max, average, standard deviation, and total revenue).

#### **SELECT**

```
System.Timestamp AS WindowEnd,
count(merchant_fee) AS CarCount,
min(merchant_fee) AS MinRev,
max(merchant_fee) AS MaxRev,
avg(merchant_fee) AS AvgRev,
stdev(merchant_fee) AS VarRev,
sum(merchant_fee) AS TotalRev
FROM SwipeStream TIMESTAMP BY swipe_date
GROUP BY TUMBLINGWINDOW(minute, 3)
```



#### **DateDiff and Time**

What is the duration between the first transaction in the window and the last transaction in the window? What was the duration between the first transaction in the window and the end of the window?

#### **SELECT**

```
System.Timestamp AS WindowEnd,
count(*) AS Frequency,
datediff(second, min(swipe_date), max(swipe_date)) AS FirstLastDuration,
datediff(second, min(swipe_date), System.Timestamp) AS FirstEndDuration
FROM SwipeStreamTIMESTAMP BY swipe_date
GROUP BY TUMBLINGWINDOW(minute, 3)
```



#### Joining Stream with Reference Data

 Say we had a list of stolen credit card numbers. Let's run each transaction against this list and get the locations.

```
SELECT
      SwipeStream.swipe date as SwipeTime,
      SwipeStream.card number as CardNumber,
      SwipeStream.merchant as Store,
      SwipeStream.swipe city state as Location,
      StoleList.Stolen as Stolen
FROM SwipeStream TIMESTAMP BY swipe date
JOIN StolenList
ON SwipeStream.card number = StolenList.card number
WHERE StolenList.Stolen = '1'
```

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## Joining Streams, Temporally

- How long did it take for each transaction to get approval from the bank?
  - Joining on events through time
  - JOIN operator requires specifying a temporal wiggle room describing an acceptable time difference between the joined events
  - If two transactions occurred within the same join interval, then consider them the same event.



## **Joining Streams**

**SELECT** 

• How long did it take for each transaction to get approval from the bank?

```
swipe.transaction_id
swipe.swipe_date,
bank.approval_time,
DATEDIFF ( second, swipe.swipe_date, bank. approval_time) AS DurationInSeconds
FROM SwipeStream AS swipe TIMESTAMP BY swipe_date
JOIN BankStream AS bank TIMESTAMP BY approval_time
ON (swipe.transaction_id = bank.transaction_id)
AND DATEDIFF ( minute, swipe, bank ) BETWEEN 0 AND 15
```

#### Joining Streams, by Window

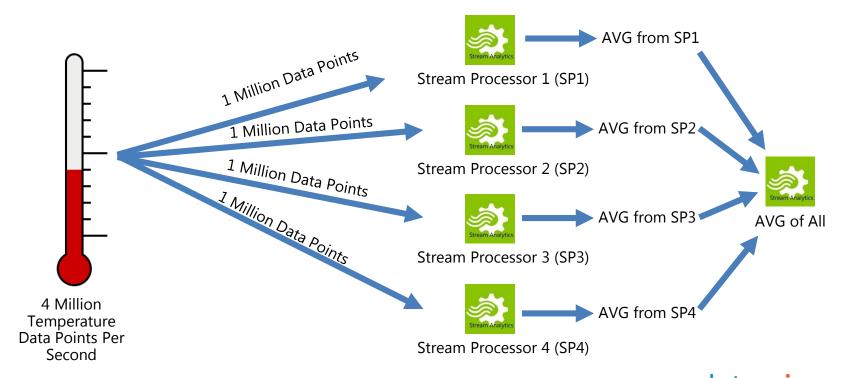
What was the average time that it took for transactions to get approved every 3 minutes?

#### **SELECT**

```
System.Timestamp AS WindowEnd,
avg( DATEDIFF ( second, en.swipe_date, ex.ExitTime )) AS DurationInMinutes
FROM SwipeStream AS swipe TIMESTAMP BY swipe_date
JOIN BankStream AS bank TIMESTAMP BY approval_time
ON (swipe.transaction_id = bank.transaction_id)
AND DATEDIFF ( minute, swipe, bank ) BETWEEN 0 AND 15
Group by TumblingWindow( minute, 3)
```



#### Average of Average Approximations





#### **Built-In Functions And Supported Types**

```
Aggregate functions
Count, Min, Max, Avg, Sum
Scalar functions
Cast
Date and time
Datename, Datepart, Day, Month, Year,
Datediff, Dateadd
String
Len, Concat, Charindex, Substring,
Patindex
```



#### QUESTIONS

