

### How NLP is Helping a European Financial Institution Enhance Customer Experience

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### About GigaSpaces

We provide one of the leading in-memory computing platforms for real-time insight to action and extreme transactional processing. With GigaSpaces, enterprises can operationalize machine learning and transactional processing to gain real-time insights on their data and act upon them in the moment.



InsightEdge is an in-memory realtime analytics platform for instant insights to action; analyzing data as it's born, enriching it with historical context, for smarter, faster decisions



In-Memory Computing Platform for microsecond scale transactional processing, data scalability, and powerful event-driven workflows





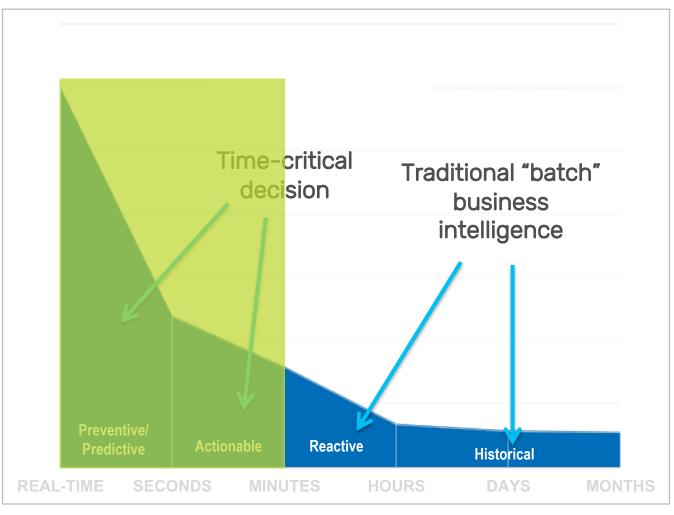
**5,000+** Large installations in production (OEM)

**25+** ISVs



### 74% want to be data driven/ only 23% are successful, **FORRESTER**<sup>®</sup>

### How Can You Gain the Most Value from Your Data?



Value

Near real-time data is highly valuable if you act on it on time

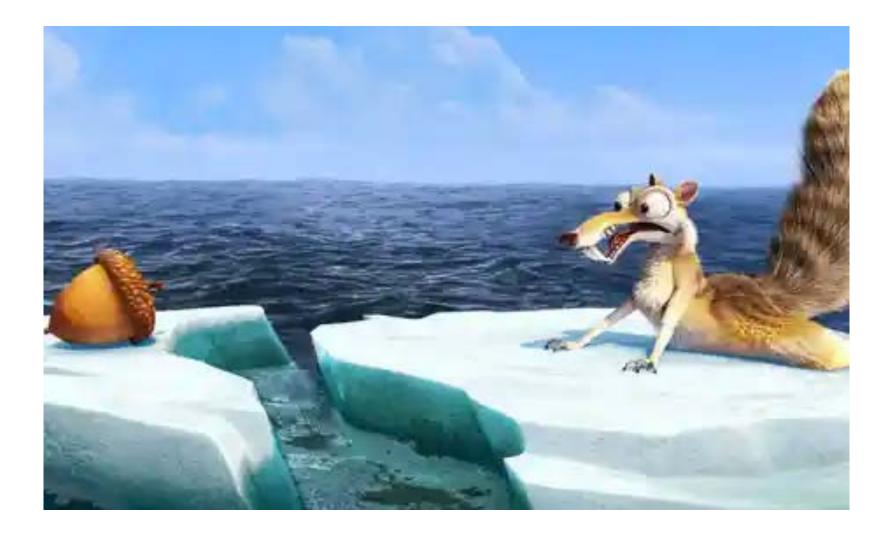
Historical + near real-time data is more valuable if you have the means to combine them

Time













### The Velocity of Business (once upon a time)

"To prevent fraud, anomaly detection needs to happen against 500,000 txn/sec in less than 200 milliseconds" "A typical e-commerce website will experience 40% bounce if it loads in more than 3 seconds, including personalization offers" "A call center receives 450,000 calls/min, across 200 phone numbers, each call needs to be routed in less than 60 milliseconds"



FINANCIAL SERVICES



ECOMMERCE



**TELCO** 



### **ABOUT THE CUSTOMER**

This Financial IT Service provider serves the leading banks in Germany with core solutions and services

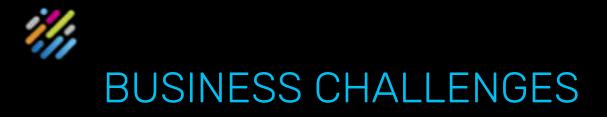
#### Business Goals:



Enhance customer experience with quicker First Call Resolution



Reduce Average Handle Time for optimized efficiency



#### DISJOINTED CUSTOMER INTERACTIONS

KEEPING UP WITH EMPOWERED CUSTOMERS

Disparate data sources and systems, led to inefficient juggling between screen and systems and poor data quality & poor customer experience Customers are smarter and have more insights into competitive products and services, raising expectations to a new standard

#### AN OMNICHANNEL EXPERIENCE

Customers want a consistent experience across all channels and agents, demanding faster resolution times



#### HIGH PERFORMANCE

MILLISECOND LATENCY

Ingestion of millions of CRM cases and data from other repositories into a unified analytics platform Customers demand an immediate response time, requiring high performance solutions that leverage ML models in real-time CONTINUOUS ML TRAINING

Insights constantly need to adapt to changing conditions for smartest insights



### **PROPOSED SOLUTION**

If a live agent is needed during a call, the NLP based solution automatically supplies the agent with articles and knowledge documents based on the conversation

#### **DATABANK**

<b>Ticket ID</b> #54367	DATA SOURCES	CUSTOMER	CUSTOMER TICKET
Customer Name #54367 Type Enterprise	International Payment declined 71.53%	International Payment declined 77.98% Beneficiary account dormant #180762	
Bronze	Case Description	Case Resolution	Case Description
Last Contact Date 20.12.18	International payment to supplier declined	Check that credit limit is not exceeded	Check here to email instructions to customer
	Read more	Read more	Email

Q Search...

🐝 GIGASPACES



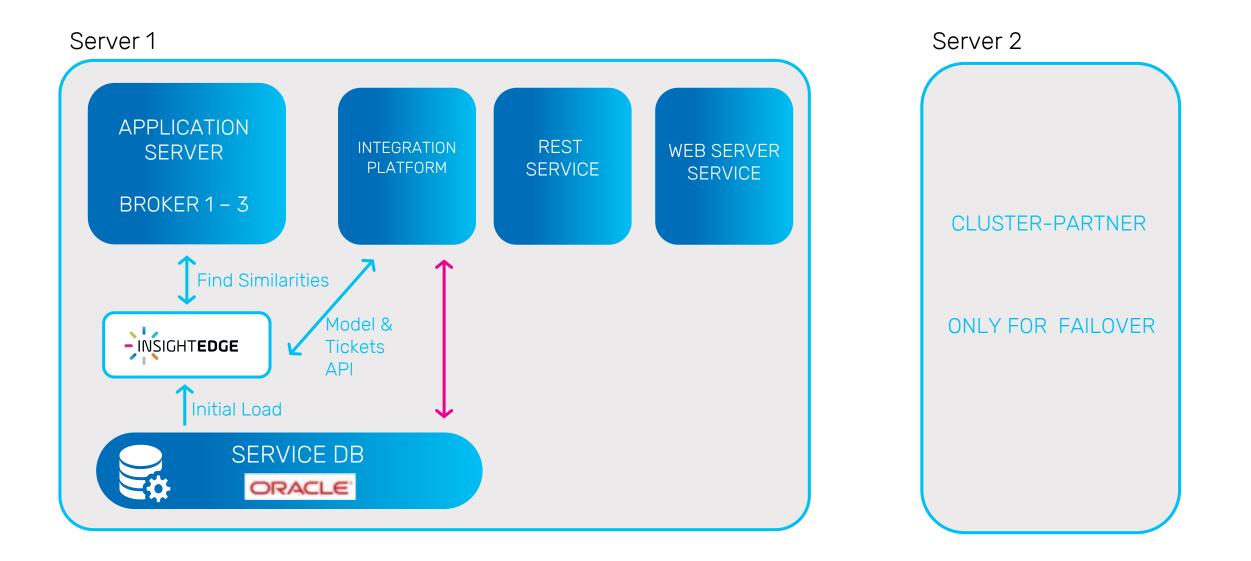
# Training the model based on

2M creater of the second state of the second s

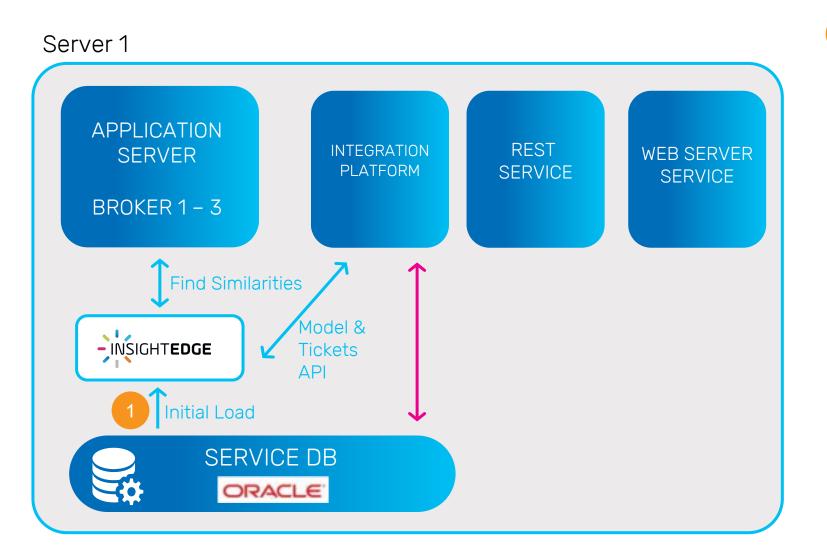
### Time to results

**50ms** 





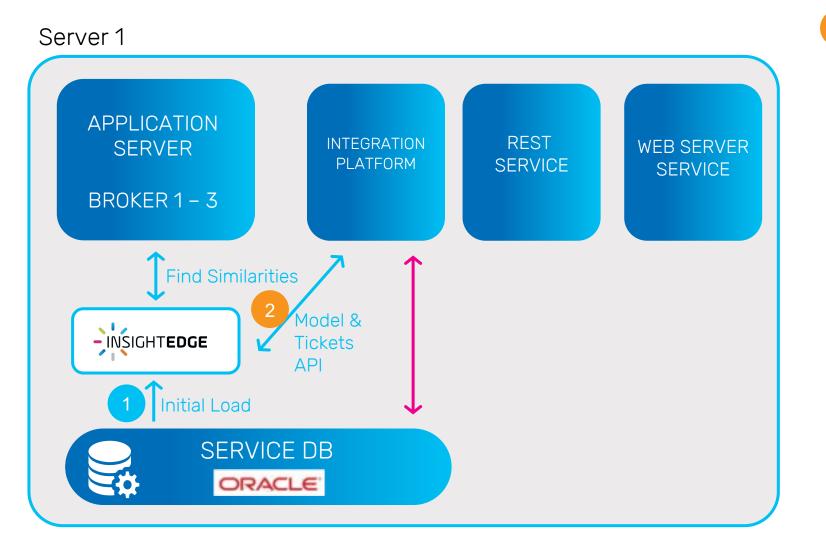




#### 1 Initial Load

Hibernate on Object Store @Id private Integer ticketID; private String searchText; private Date openDate; private Integer responsibleCompany; private Integer customerDomain; private Integer customerCompany; private Integer ticketCreator; private Integer addModelControl1; private Integer addModelControl2; private Integer addModelControl3;

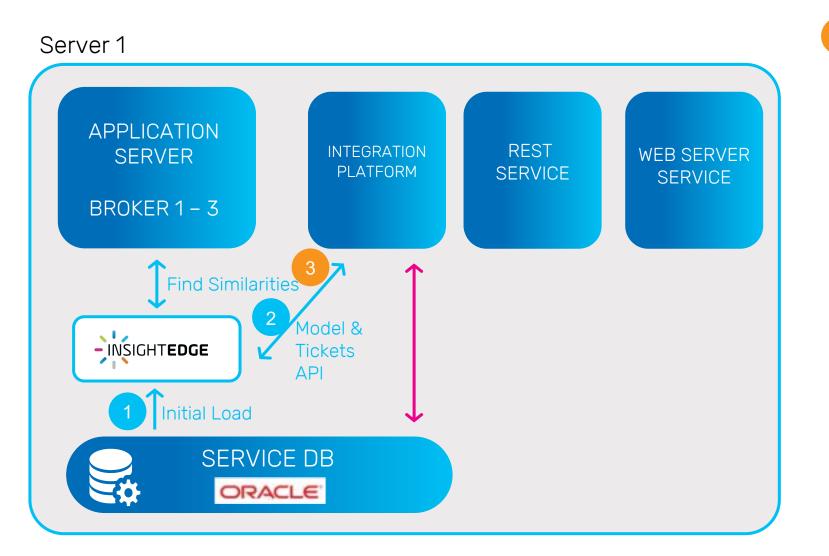




#### 2 Training/building model

- Train
- stopTrainModel
- getTrainModelStatus
- checkModelInSpace
- destroylModel

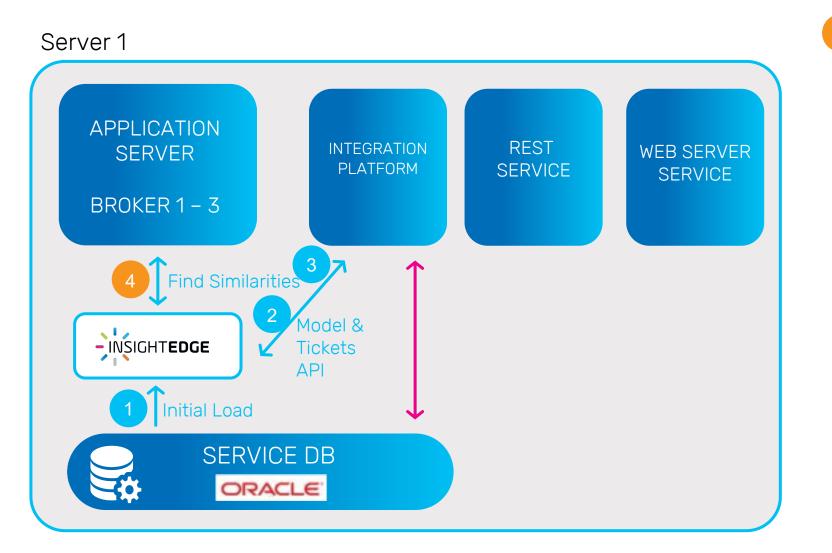




<sup>3</sup> Long Running Spark Job API

- startModel
- stopModel
- checkModellsRunning
- getFindSimilartiesStatus

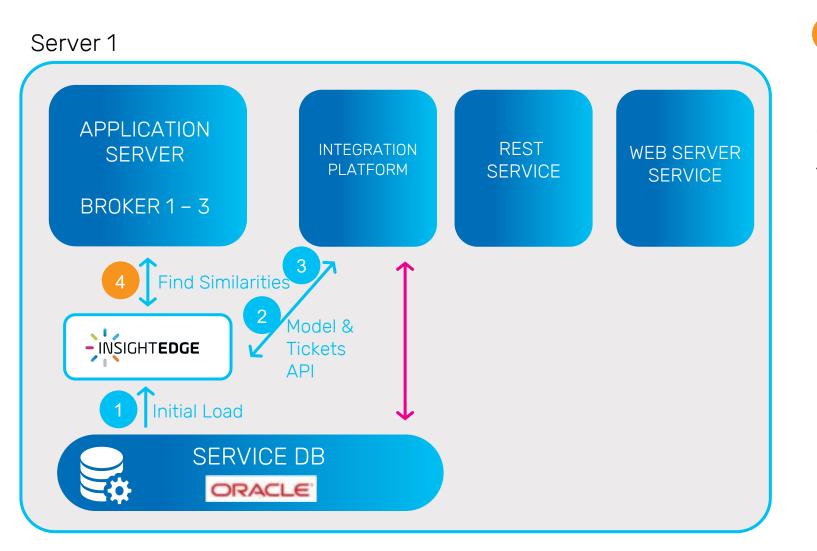




#### 4 findSimilarities

- Write findSimilaritiesRequest object to the space using task
- Spark long time running job takes the object perform the find similarities action (set the object status to processed true)

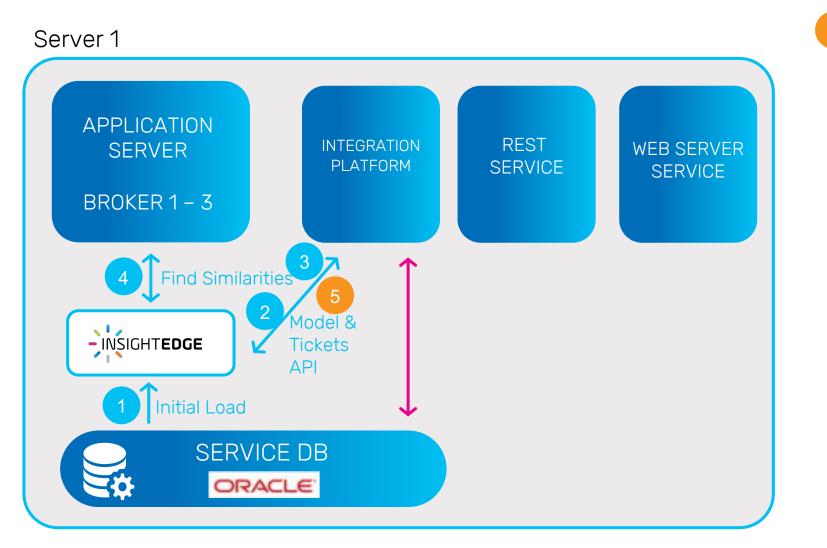




#### 4 findSimilarities

ticketId>72018 gs.exec(modelId, "my search") The result is the following similar cases: 70534 (0.823432215) 70874 (0.726937532) 70110 (0.719002341) 70998 (0.528010191)

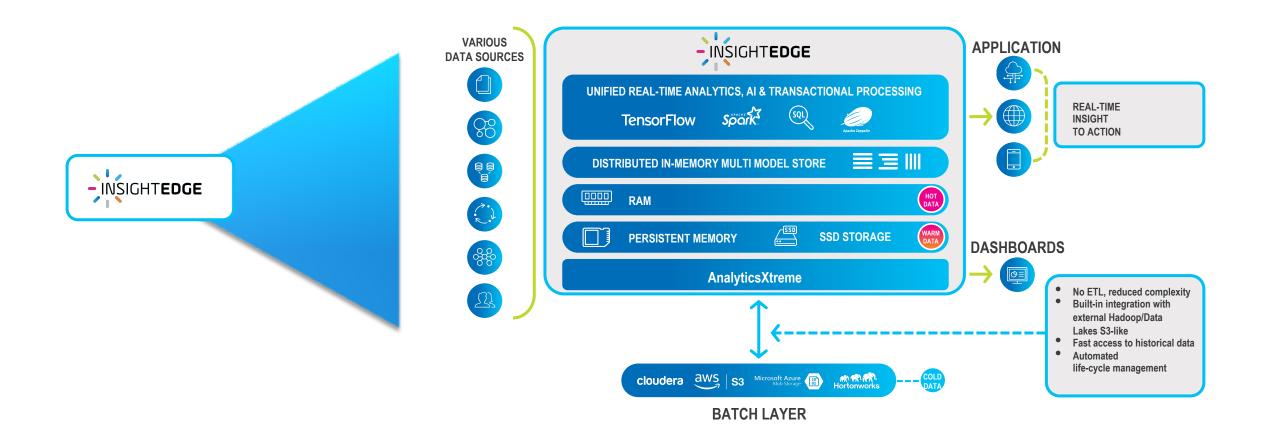




#### 5 Support Tickets (the data)

- Incremental Feed
- Delete

### Unified Transactional & Analytical Processing for Operationalizing ML





REAL-TIME	EMPOWER THE AGENT	CONTINUOUS ML TRAINING
Average time of	Allow the agents an	<b>27 Minutes</b>
50ms	immediate response	background training
to search and find	time, reducing mean	time for 2 million
similar cases	time to resolution	records

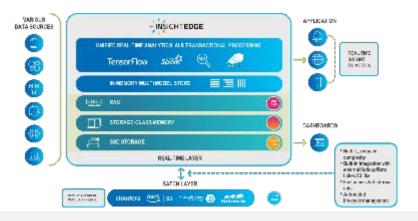


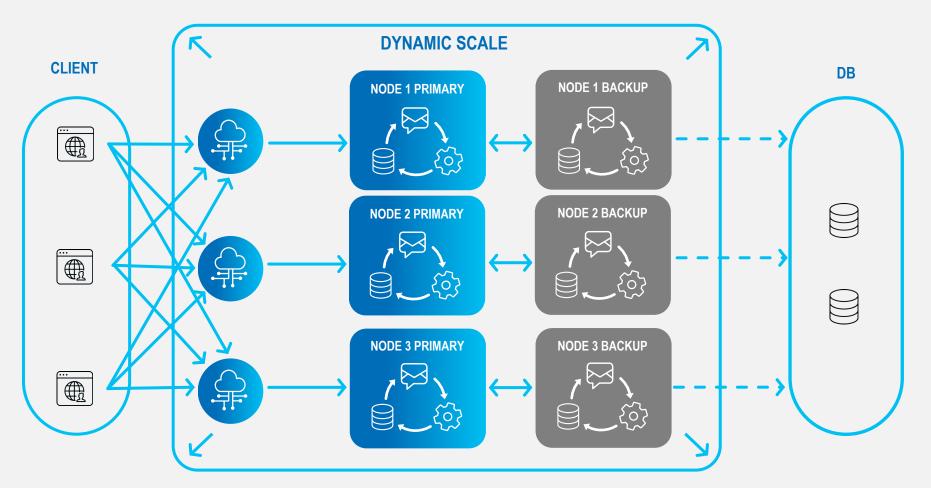
### Overcoming Challenges



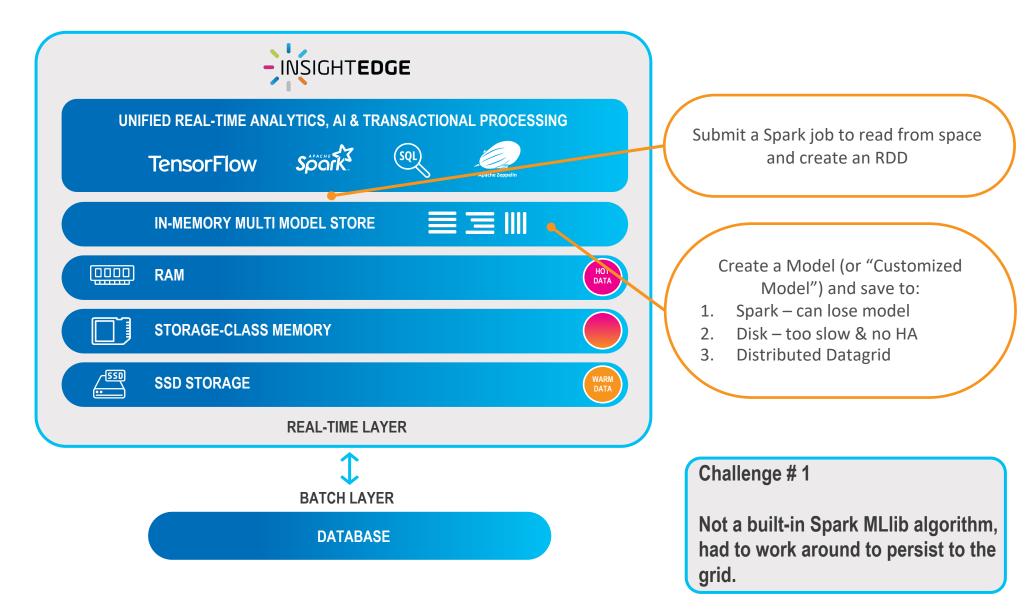
- INSIGHTEDGE	
UNIFIED REAL-TIME ANALYTICS, AI & TRANSACTIONAL PROCESSING TensorFlow	
IN-MEMORY MULTI MODEL STORE	
STORAGE-CLASS MEMORY	
SSD STORAGE	Load 2 million records from a slow tier to a distributed in-memory data fabric
REAL-TIME LAYER	(e.g. Multi-model Store)
D BATCH LAYER	
DATABASE	



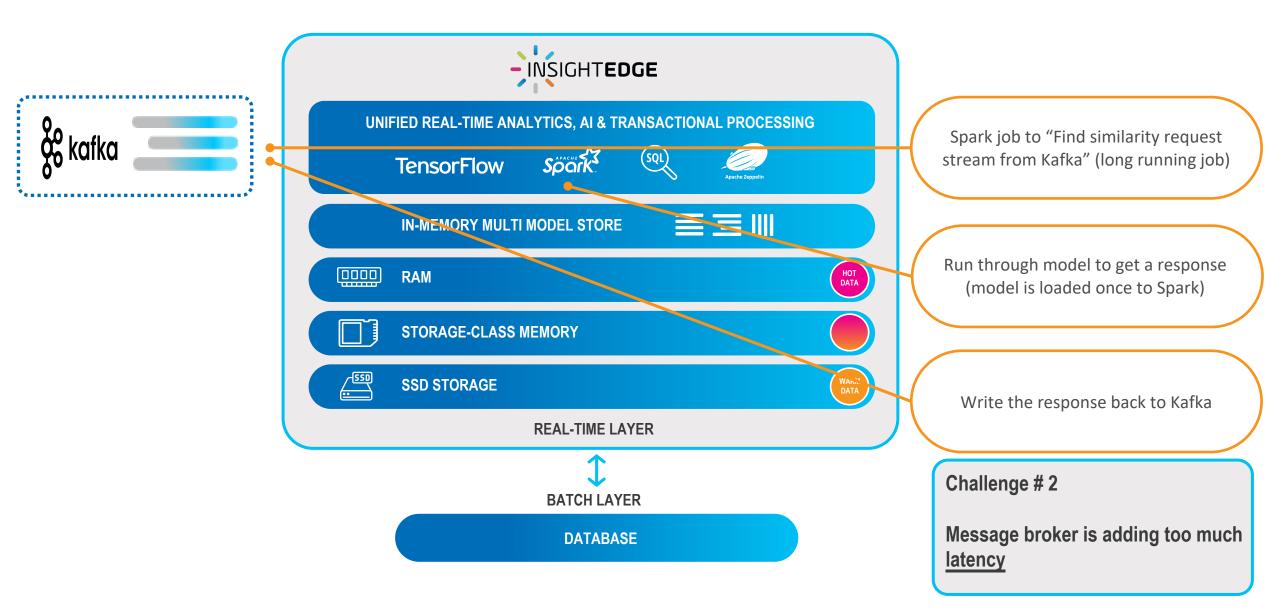




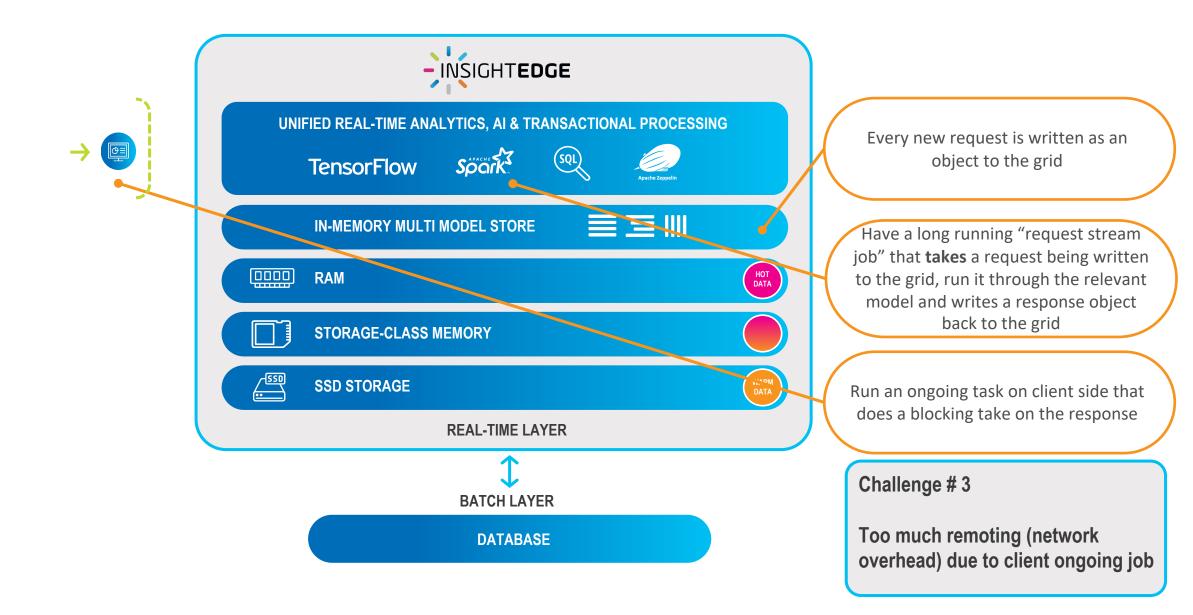
### Step 2: Create Model and Save to...



### Step 3: Request/Response via Message Broker



### Step 4: Remove Message Broker



### Step 5: Remove Remoting (as much as possible)

	- INSIGHTEDGE	
UN	IFIED REAL-TIME ANALYTICS, AI & TRANSACTIONAL PROCESSIN TensorFlow	NG We've taken the "client side code" and wrapped it up within a Grid Task (Stateless service) and deployed to the Grid
	RAM	Added the ability to route the task to different partitions if a customized model is used to reduce grid overhead
	STORAGE-CLASS MEMORY	
SSD 	SSD STORAGE	WARM DATA
	REAL-TIME LAYER	
	<b>D</b> BATCH LAYER	Challenge # 4
	DATABASE	Still too much remoting

### Step 6: Remove Remoting (cont.)

	- INSIGHTEDGE	
	ED REAL-TIME ANALYTICS, AI & TRANSACTIONAL PROCESSING	Run the job directly from the processing unit (stateful service) to further avoid remoting
IN		
	RAM	
s.	STORAGE-CLASS MEMORY	
550 St	SSD STORAGE	
	REAL-TIME LAYER	
	<b>D</b> BATCH LAYER	
	DATABASE	

### Step 7: Add Production Grade Capabilities to Spark

-INSIGHTEDGE	
UNIFIED REAL-TIME ANALYTICS, AI & TRANSACTIONAL PROCESSING TensorFlow	Use Remote Service to start the "train model" job and "start long running request stream" job
IN-MEMORY MULTI MODEL STORE	
STORAGE-CLASS MEMORY	
SSD STORAGE	
REAL-TIME LAYER	
<b>C</b> BATCH LAYER	Solution
DATABASE	Add monitoring and auto-recovery to the Spark job

### MOVING FORWARD FOCUS ON AUTOMATION

Automate Call Routing (using Deep Learning Approach)

Automatic topic determination based on text classification and sentiment analysis

Automatic CRM case creation



### ABOUT THE USE CASE

This use case shows how to modernize existing software architecture for an efficient call center routing workflow

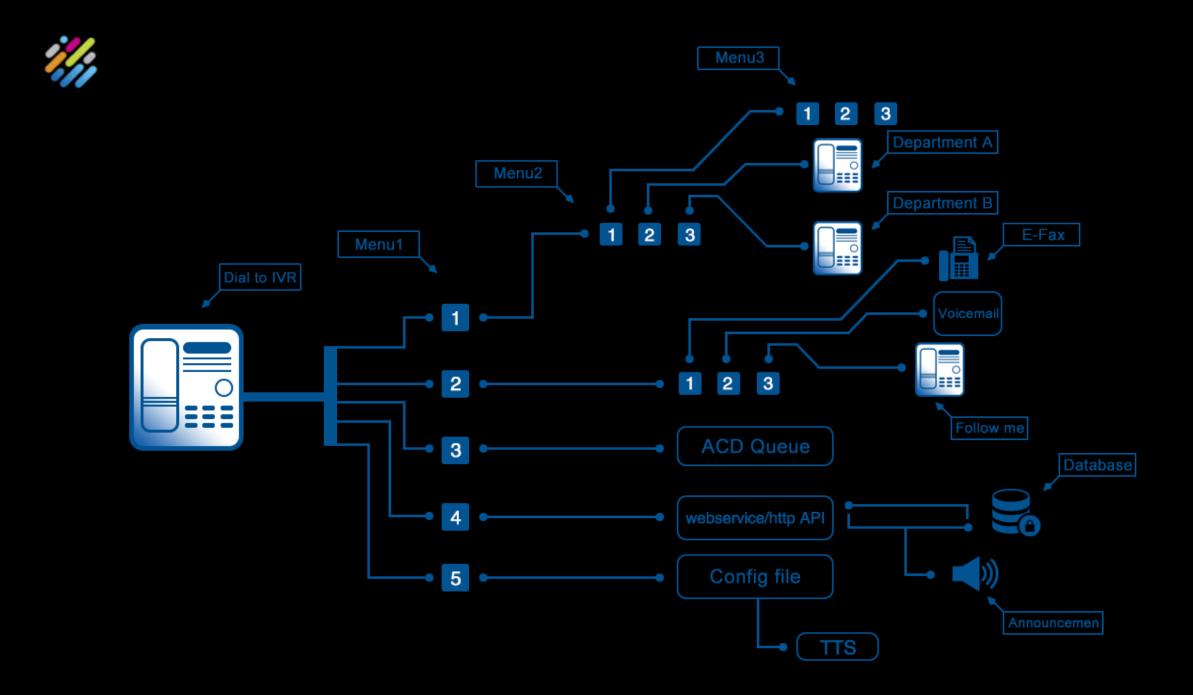
#### **JSE CASE BENEFITS:**



Enhance Customer Experience with automatic routing that prevents customers from being buried in a hierarchical menu



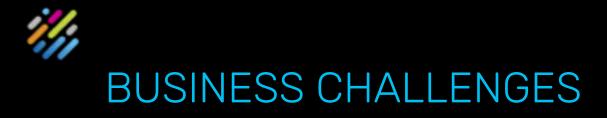
Reduce Average Handle Time for optimized efficiency





## 60%

# of callers bypass IVR for voice (costs are 12x higher because of this)



Improve Customer Experience

Reduce Costs: lower AHT

Faster call routing to the correct agent means more satisfied customers Faster call resolution: Faster routing

Routing to correct agent

+

Enhanced System Agility

Higher agility when adding new categories or departments

# TECHNICAL CHALLENGES

#### Performance

#### Simplification

#### Event Driven Architecture based on prediction criteria is required for optimal performance supporting peak events

Leveraging existing opensource frameworks such as BigDL in a unified platform simplifies architectural complexity

#### **Continuous ML Training**

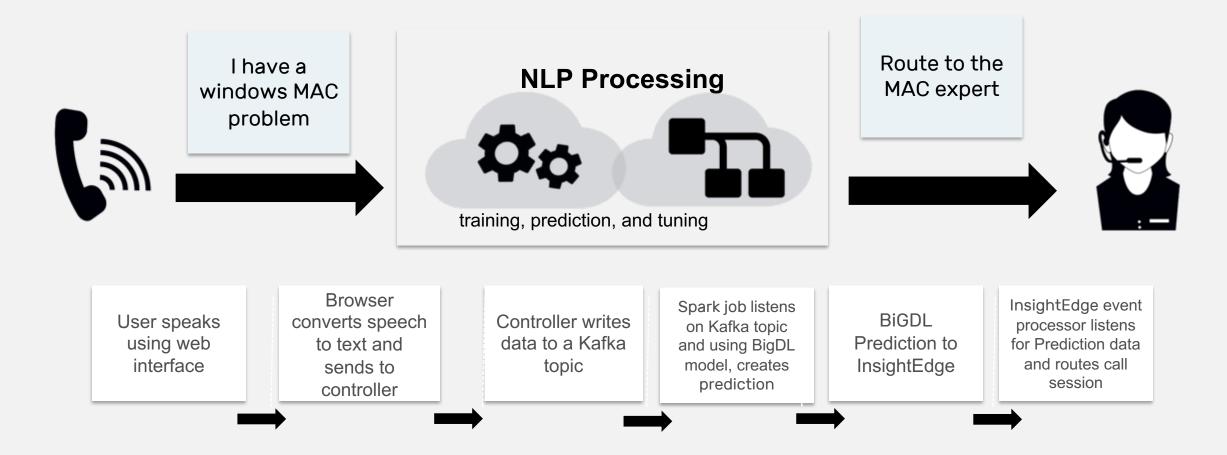
Continuous model training

based on previous transcribed calls + automatic training of alternative models ensure models with higher scoring



DEEP LEARNING InsightEdge

Automatic routing to the right agent for the perfect personalized experience



### Automatic Call Routing – Live Demo

#### Call session assistant

Search...



See the brain behind it!

Insightedge Web-UI

Spark Jobs

#### Model: TextClassification Training time: 9 min

Accuracy: 0.7778865

BigDL v0.2.0

#### Call Center BigDL/InsightEdge module

Hello it's almost evening and my Windows computer stopped working

#### In-process calls (Powered by Intel BigDL) 0

Text

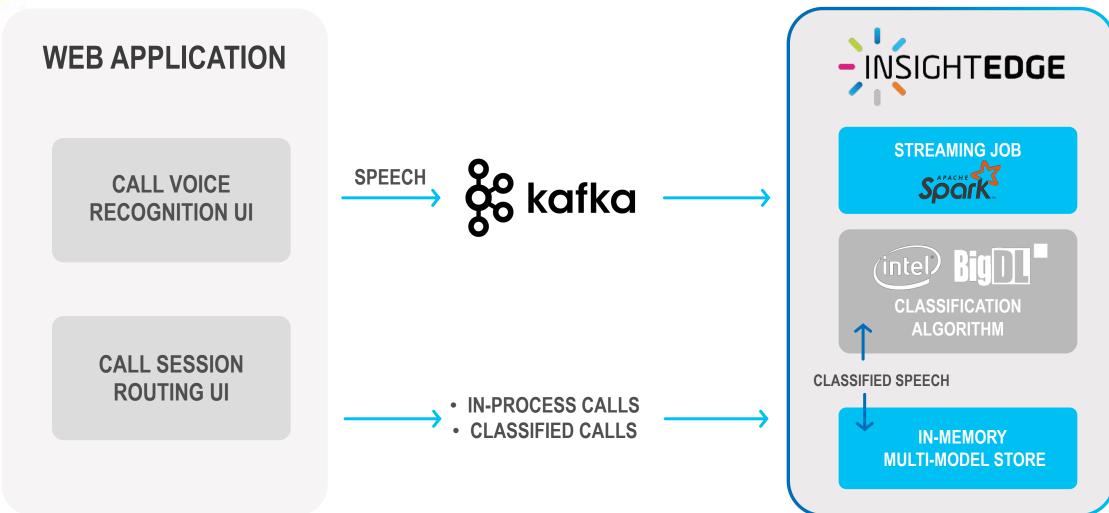
#### Call sessions 3

Id

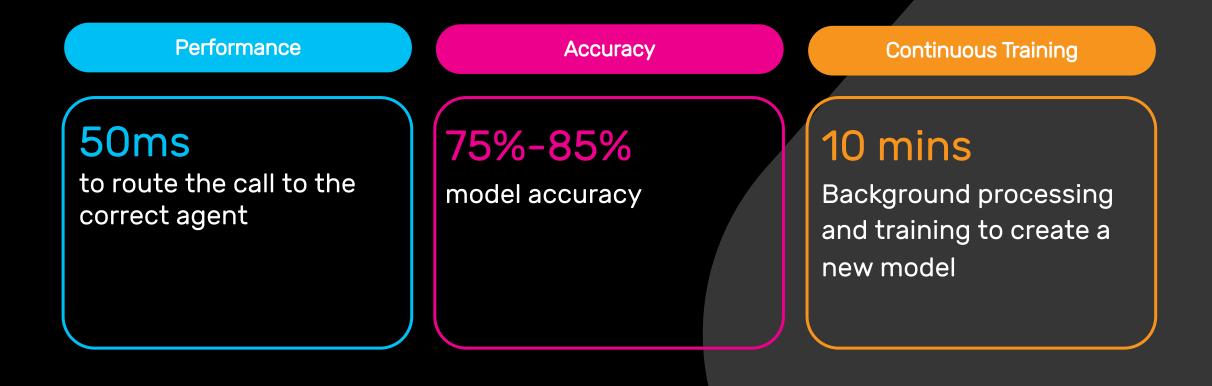
Id	Category	Agent Id	Time (ms)	Text
3	comp.os.ms-windows.misc	4	36	Hello it's almost evening and my Windows computer stopped working
2	comp.sys.mac.hardware	4	59	Hello it's a sunny day but my MacBook screen now just been broken
1	comp.sys.mac.hardware	1	86	Hello I have a problem with my Mac

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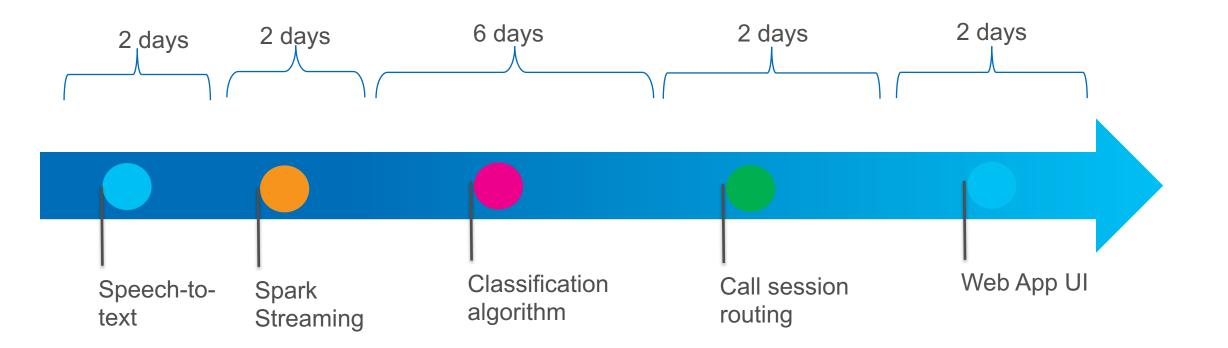








#### **Timeline to implement** end-to-end in 14 days



### Real-time Insights and Actions Require High Performance

- Unifies analytics, AI and real-time transactions
- Triggers transactional workflows based on prediction criteria and scoring
- Efficient scale-out computing
- Distributed model training
- Lowers TCO/Decreases Deployment Costs train and run large-scale deep learning workloads
- High performance



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