

KSB 인공지능 프레임워크

2018. 9. 13

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KSB Convergence Research Department, **ETRI**

Agenda

- I. Issus in Developing AI Services**
- II. Recent trend in AI Platforms**
- III. Strategies of KSB Framework**
- IV. Usecases**
- V. Conclusion**

I. Issus in Developing AI Services

AI Opportunity and Challenges

AI Opportunity:

Opportunity to transform their business by implementing sophisticated models for recommendation engines, ads targeting, speech recognition, object recognition, bots, sentiment analysis, predictive analysis, and more.

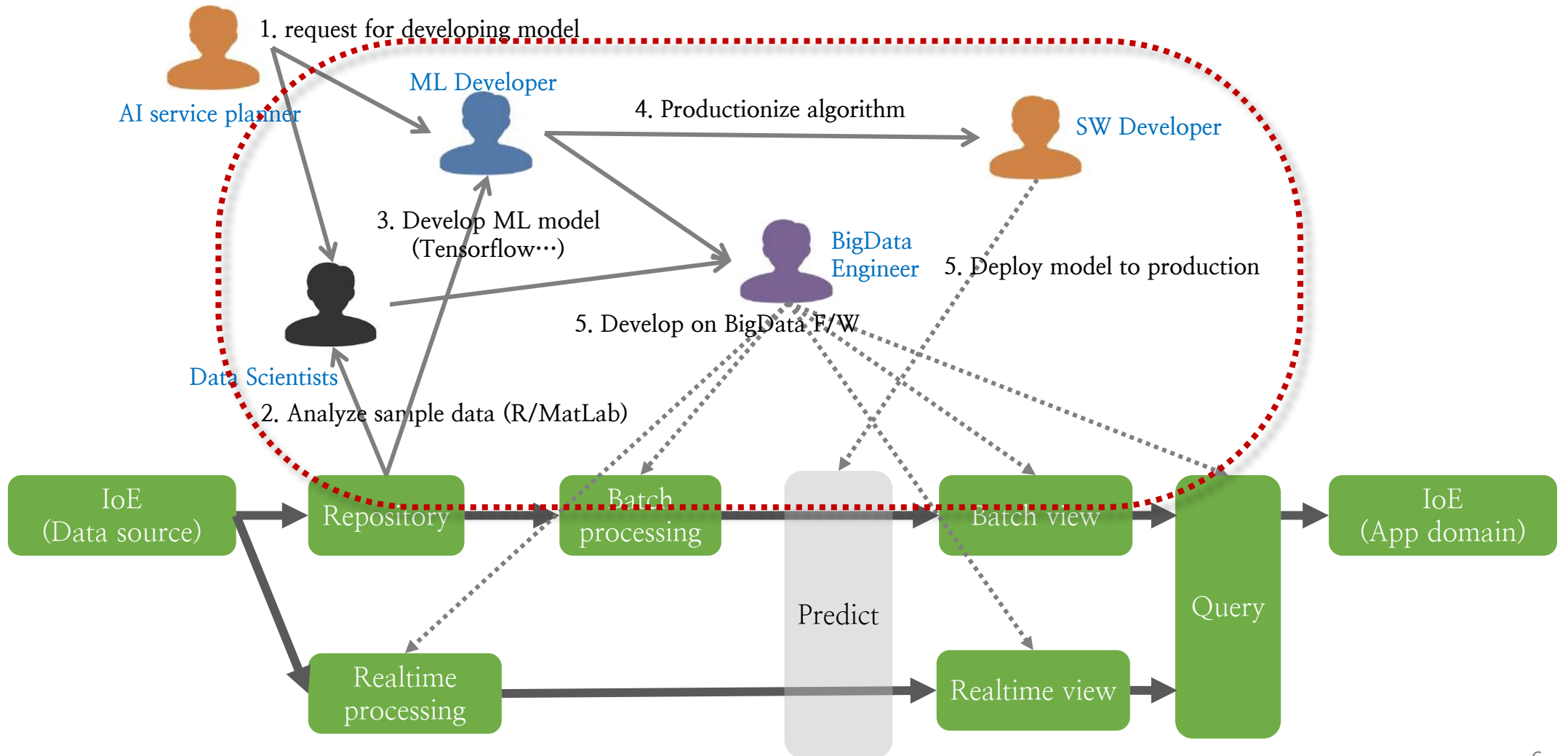
However, practitioners face multiple challenges when implementing such applications:

- **Environment Configuration:** Ability to setup and maintain complex environments due to the multitude of Machine Learning frameworks available.
- **Distributed Training:** Ability to train models in a distributed fashion to get results faster. This requires specialized skills and complex code to manage.
- **Compute Power:** Ability to run parallelized workloads on GPUs for maximum performances. This is essential for Deep Learning applications.

Era of AI, but still there are many difficulties ...

- Diverse AI frameworks
 - Diverse ML/DL libraries – Tensorflow, MXNet, PyTorch, Caffe2, Torch, theano
- Diverse form of data to the model
- Deliver data to the model on time
- Update model for understanding recent trends
- Interact from models to model for solving one problem
- Cloud dependency
- ...

Journey from Data to AI Application



Machine Learning in Real World

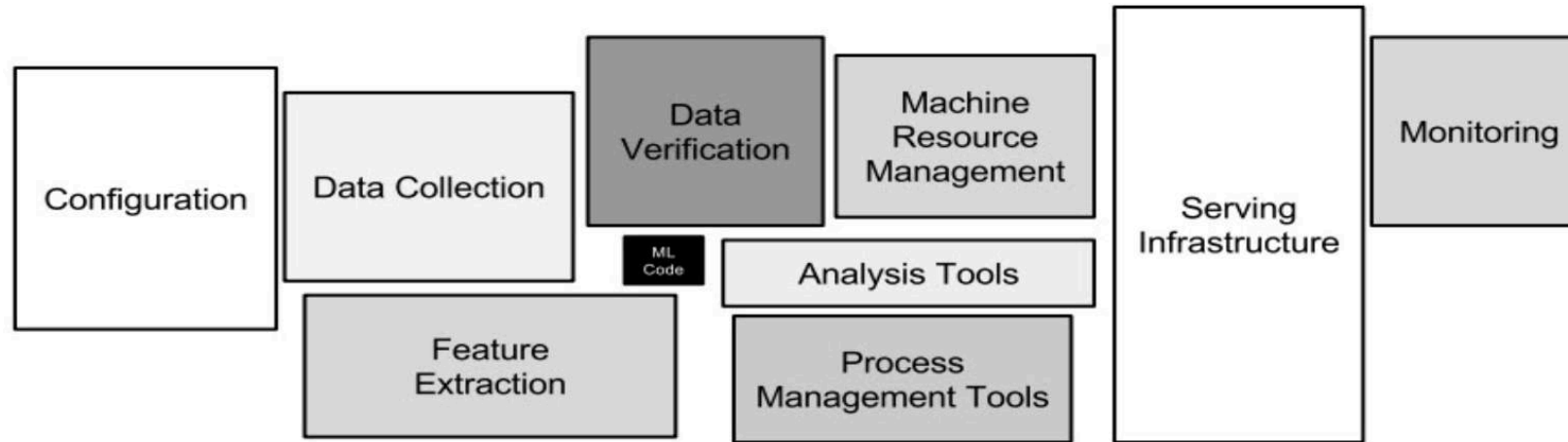
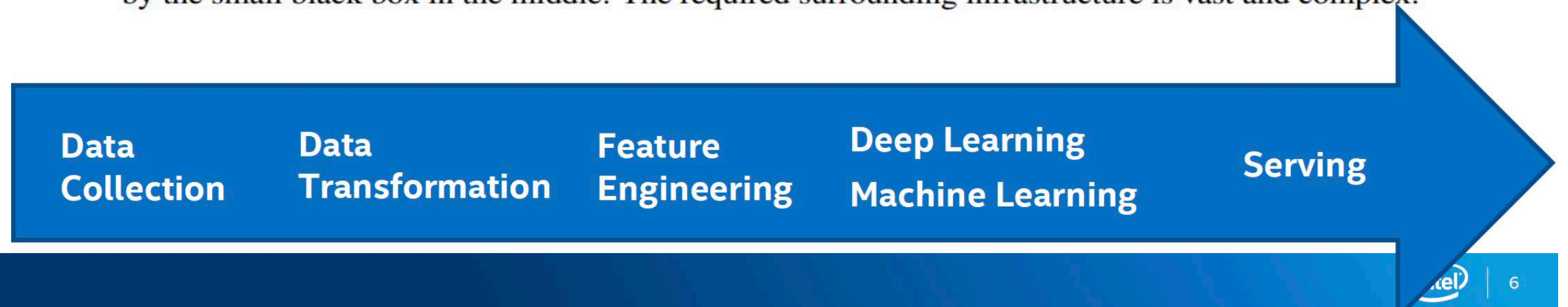


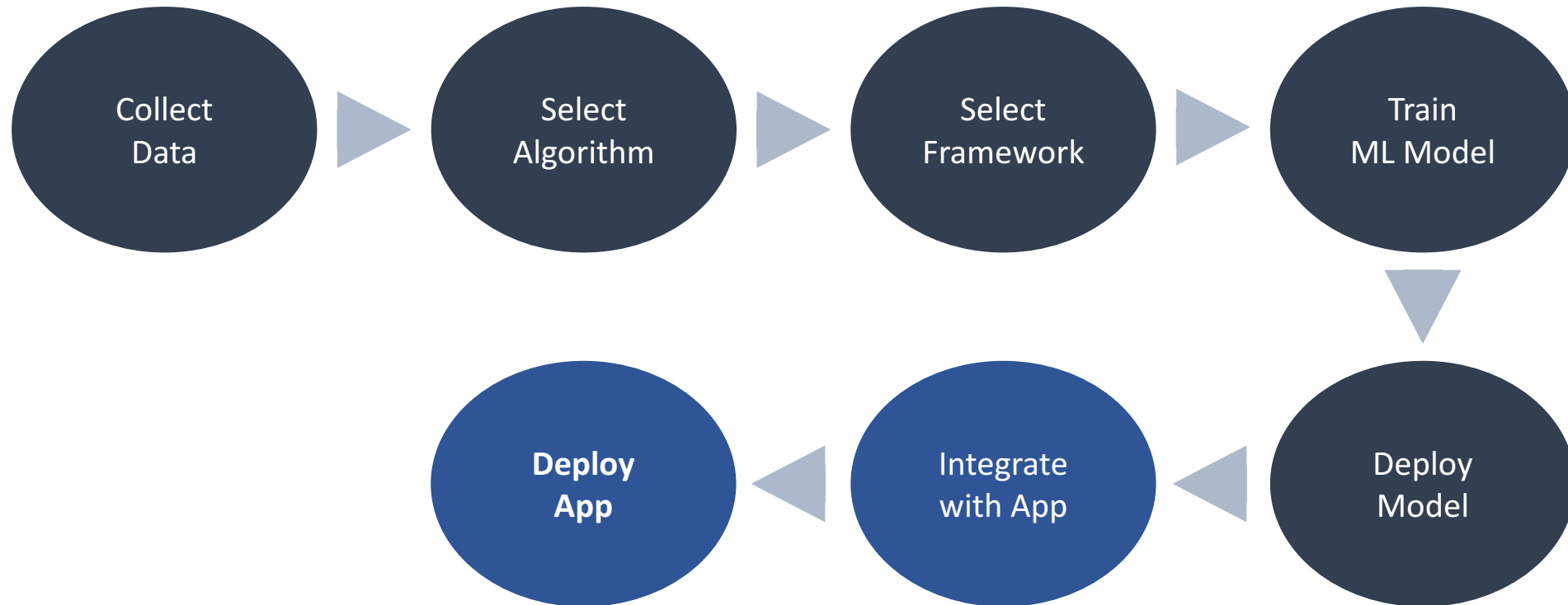
Figure 1: Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small black box in the middle. The required surrounding infrastructure is vast and complex.



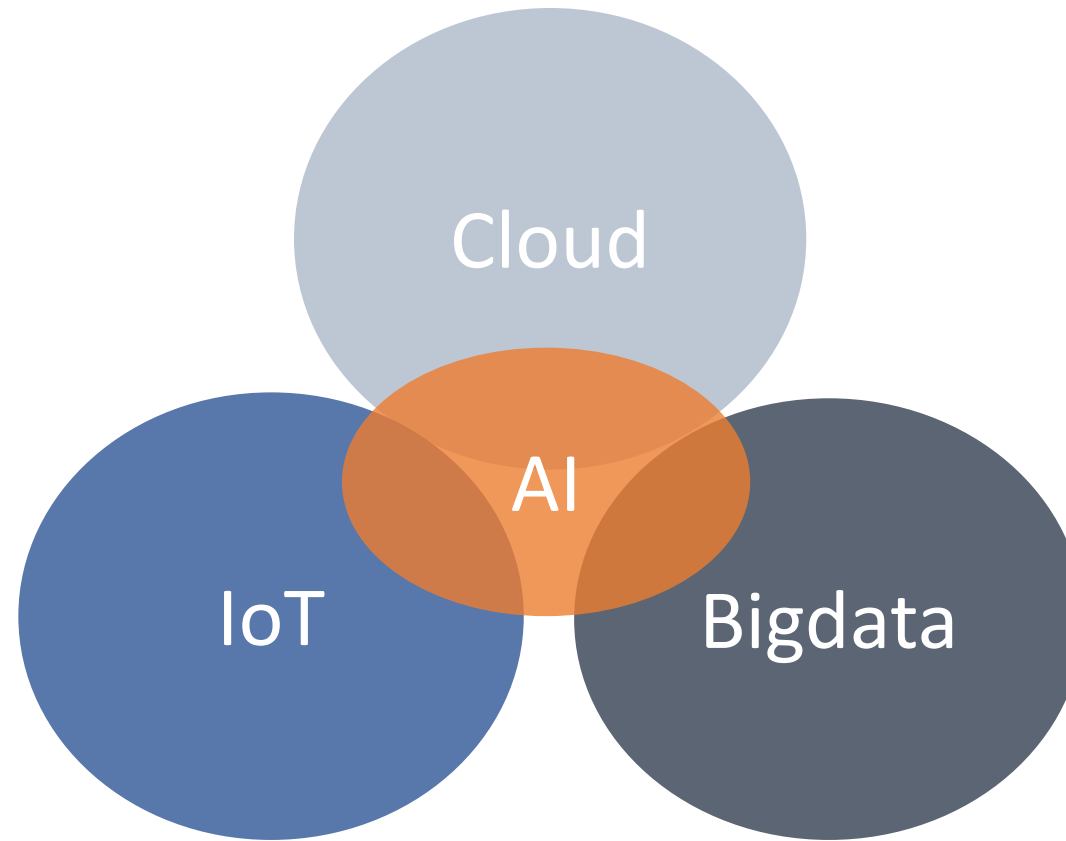
Source: Future of data analytics (Intel, Open Data Science Conference 2017)

II. Recent trend in AI Platforms

Journey from Data to AI Application

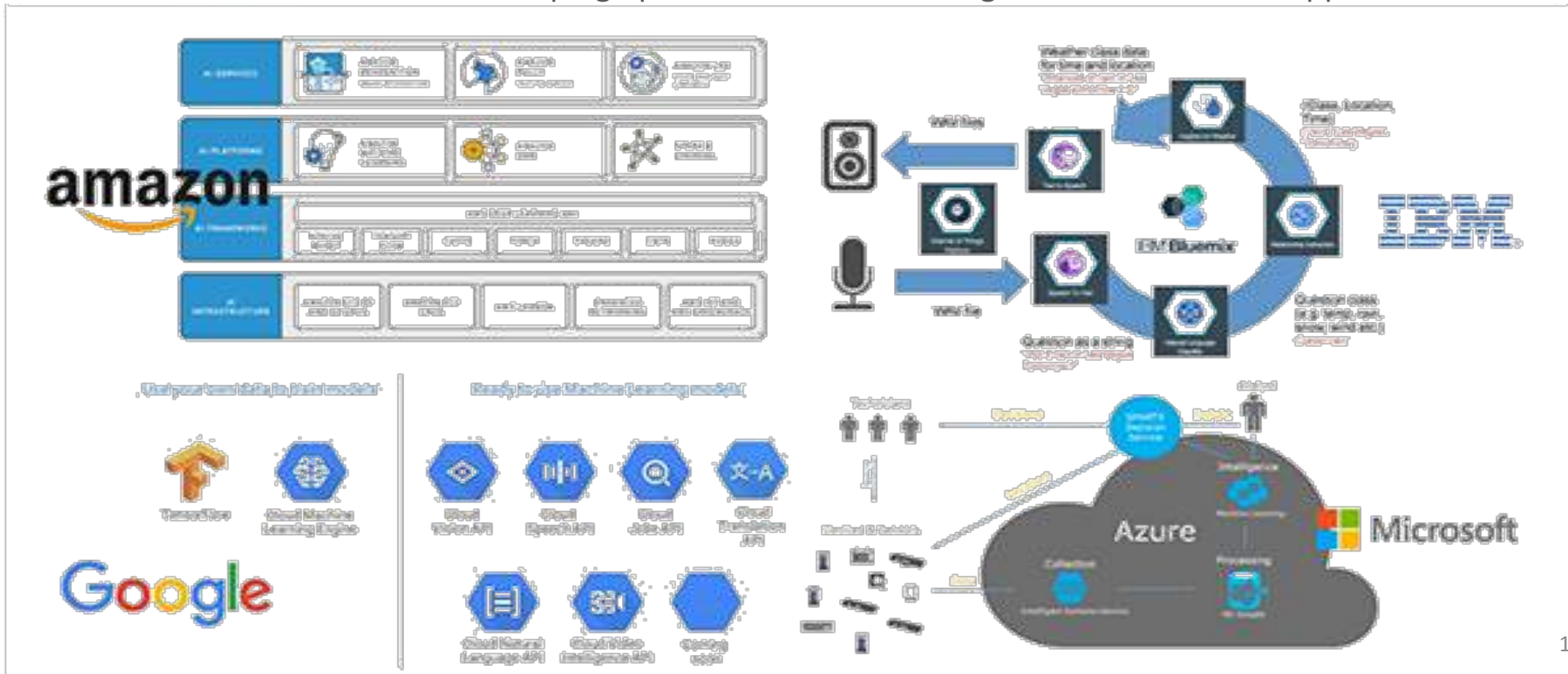


Global IT Companies

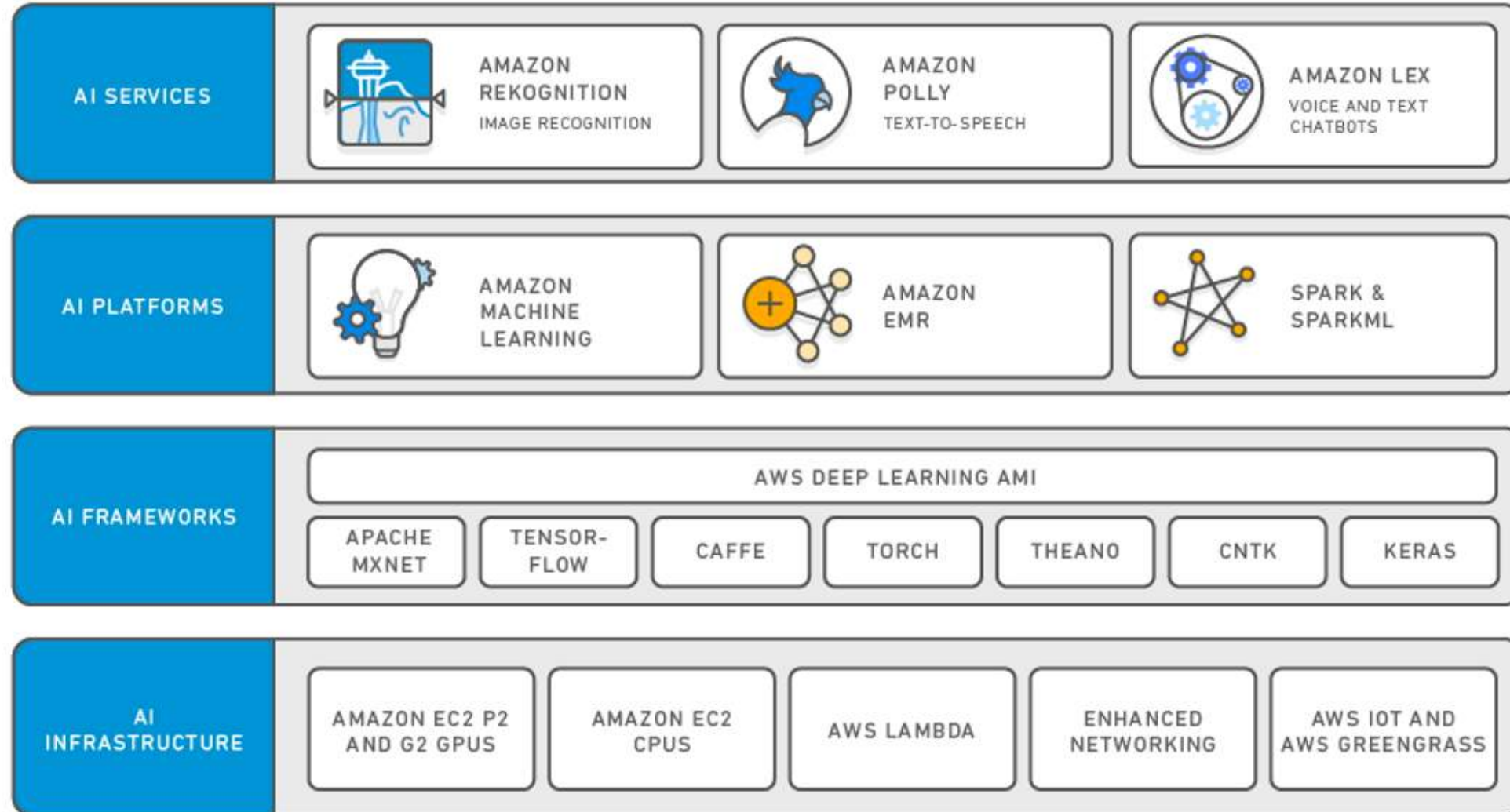


Cloud + AI

- Global IT Companies (Google, Amazon, IBM, MS)
 - Provide cloud computing-based cognitive service and ML service platform
 - Provide foundation for developing open source-based intelligent services and IoT application



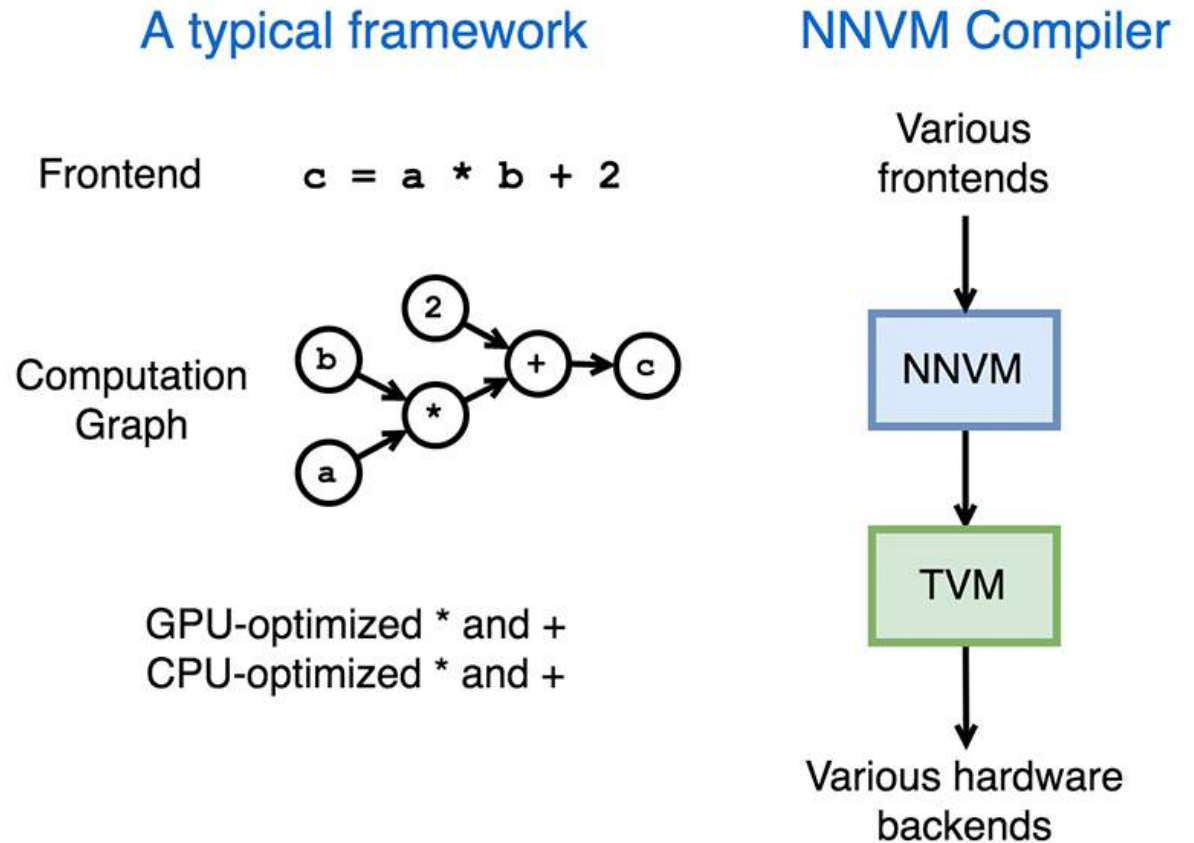
Cloud + AI: Amazon AI



Cloud + AI: Amazon AI

NNVM Compiler: Open Compiler for AI Frameworks

- Three challenges:
 - Switch from one AI framework
 - (e.g. pyTorch -> Caffe2)
 - Multiple backends to guarantee performance
 - Supporting multiple frameworks requires enormous engineering efforts



Source: <https://aws.amazon.com/ko/blogs/machine-learning/introducing-nnvm-compiler-a-new-open-end-to-end-compiler-for-ai-frameworks/>

Cloud + AI: Amazon AI

NNVM Compiler: Open Compiler for AI Frameworks

- an end-to-end compiler based on the TVM stack that compiles workloads directly from various deep learning frontends into optimized machine codes.

Typical Workflow of NNVM Compiler

model from framework

```
graph, params = nnvm.frontend.from_xyz(...)
```

compile

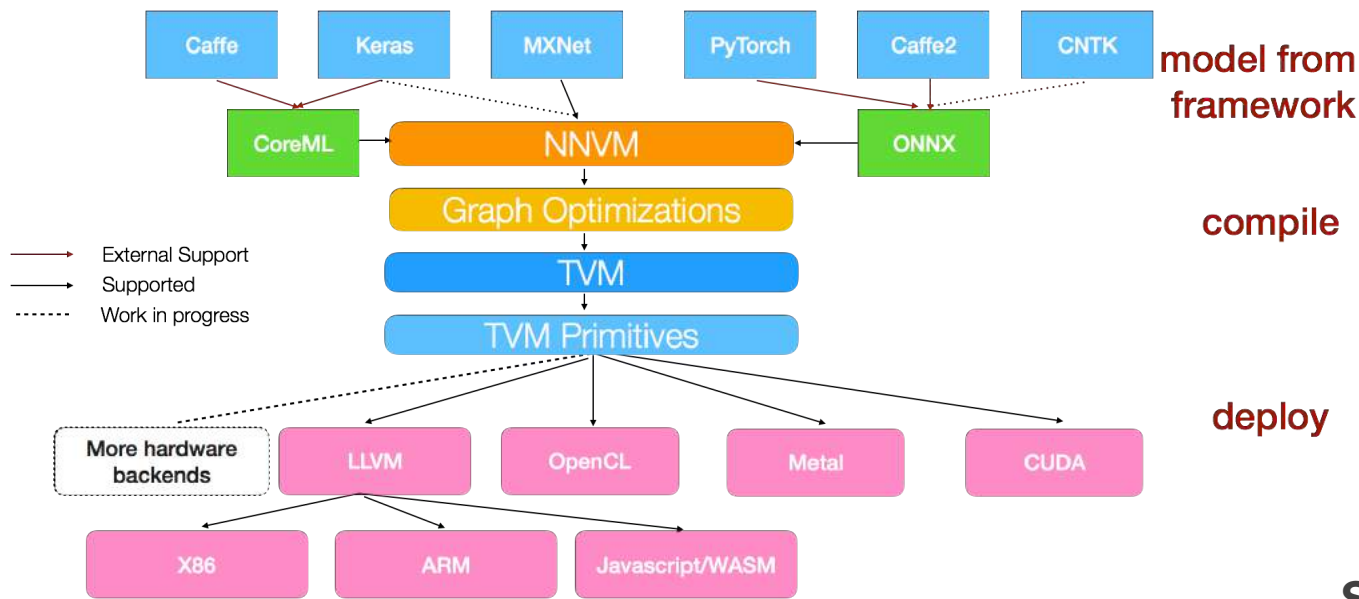
```
graph, lib, params = nnvm.compiler.build(
    graph, target="cuda", {"data", data_shape}, params=params)
```

deploy

```
module = graph_runtime.create(graph, lib, tvm.gpu(0))
module.set_input(**params)
module.run(data=data_array)
output = tvm.nd.empty(out_shape, ctx=tvm.gpu(0))
module.get_output(0, output)
```

Separation of Optimization and Deployment

* Source: NNVM Compiler: Open Compiler for AI Frameworks



Cloud + AI: Google AI

Use your own data to train models



TensorFlow

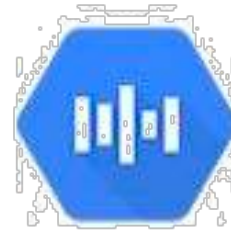


Cloud Machine Learning Engine

Ready to use Machine Learning models



Cloud Vision API



Cloud Speech API



Cloud Jobs API



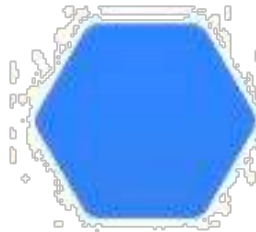
Cloud Translation API



Cloud Natural Language API



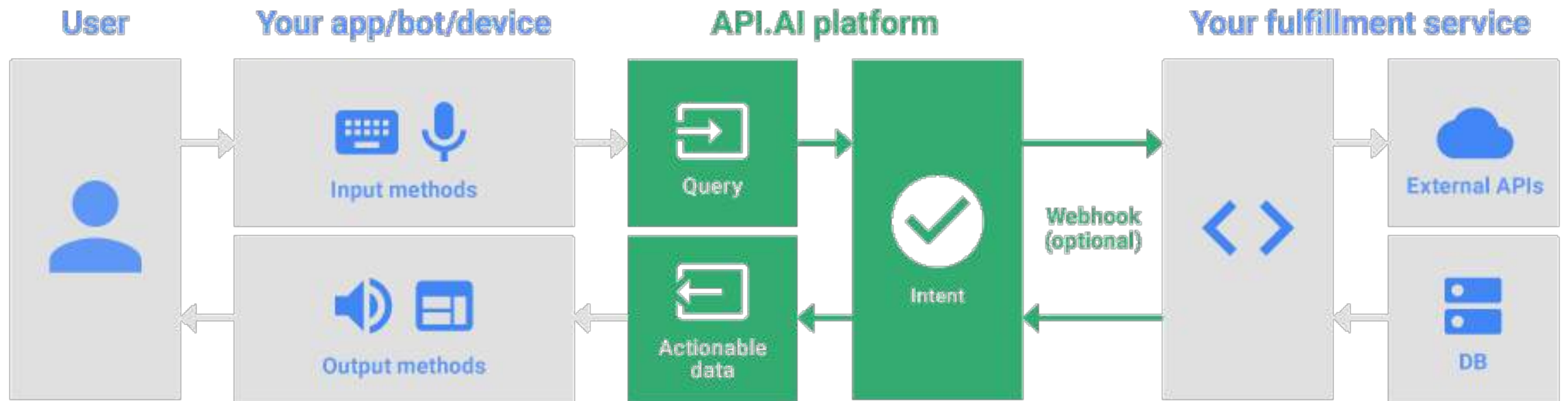
Cloud Video Intelligence API



Coming soon

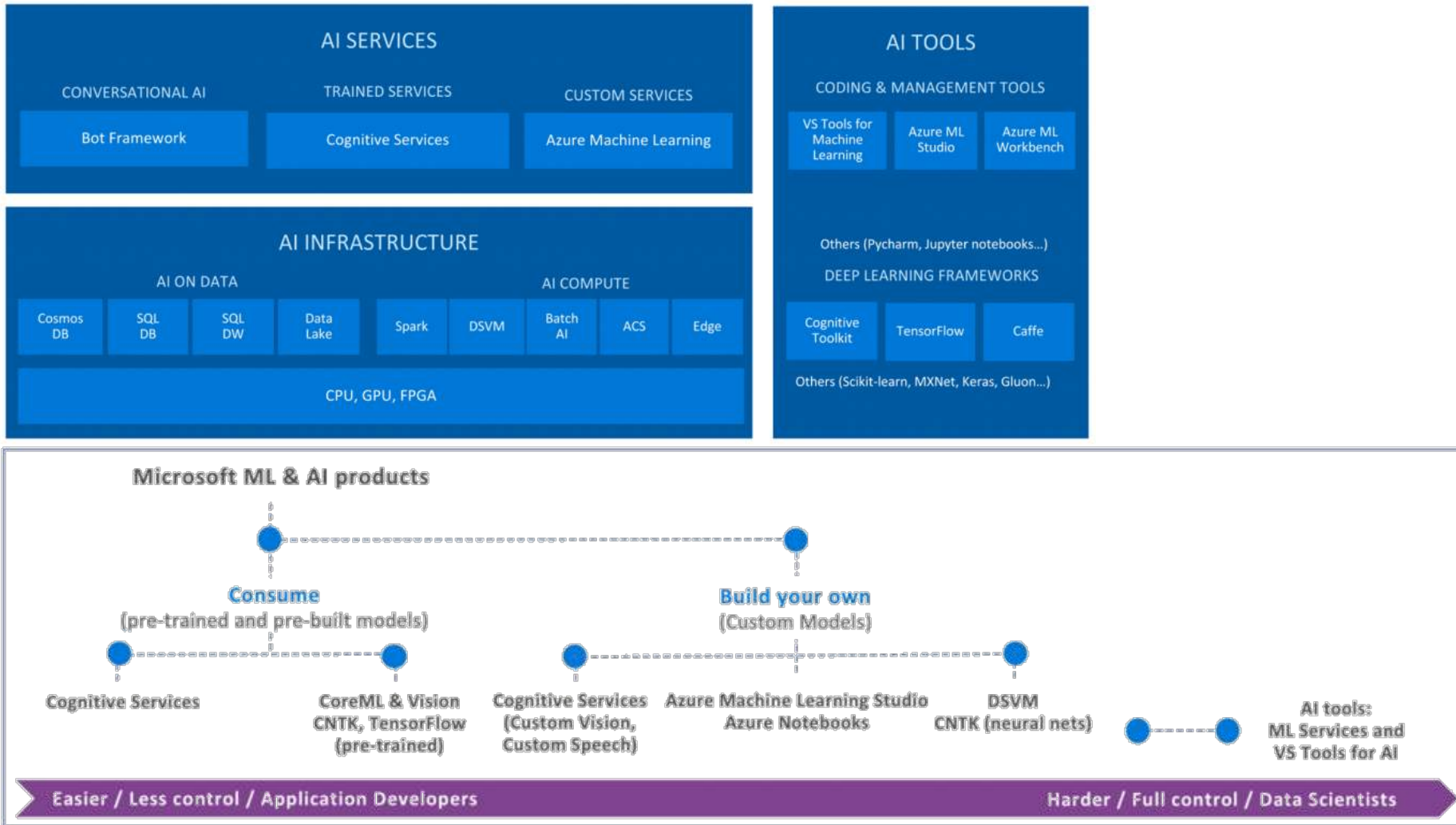
Cloud + AI: Google AI

Dialogflow: Build natural and rich conversational experiences



* source: <https://dialogflow.com>

Cloud + AI: Microsoft AI



Cloud + AI: IBM AI

IBM PowerAI Platform

PowerAI Software Distribution

Deep Learning Frameworks

Caffe



Caffe

IBM Caffe



torch



theano



Chainer

Supporting Libraries

DIGITS

OpenBLAS

Distributed Frameworks

Bazel

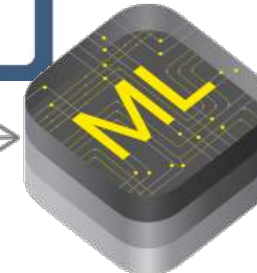
NCCL

IBM Power System for HPC, with NVLink

Breakthrough performance for GPU accelerated applications, Including Deep Learning and Machine Learning



Core ML model



Core ML



Your app

Build a production-ready iOS app

BigData + AI



BigData + AI: Databricks

Open source project announcements:

- **MLflow:** MLflow (currently in alpha) is a cross-cloud open source framework designed to manage the entire machine learning lifecycle and work with any machine learning library.
- **Databricks Runtime for ML:** Based on customer demand, Databricks announced the new native and deep integration of popular ML libraries as part of the Databricks Runtime, including xgboost, scikit-learn, numpy as well as TensorFlow, Keras and Horovod. You can find more information [here](#).
- **Analytics-ready Data with Databricks Delta:** Simplify data reliability and performance of Apache Spark™ with Databricks Delta. Ensure your data is ready for analytics. [Watch the keynote by Apple](#) and demo to learn more.
- **Unified Analytics Platform for Genomics:** Accelerate discovery with a collaborative platform for genomic data processing, tertiary analytics and AI at massive scale. Read the blog and sign-up for our private preview.

BigData + AI: Databricks

AI Opportunity vs. Challenges

AI Opportunity:

Today more than ever, data scientists and Machine Learning practitioners have the opportunity to transform their business by implementing sophisticated models for recommendation engines, ads targeting, speech recognition, object recognition, bots, sentiment analysis, predictive analysis, and more.

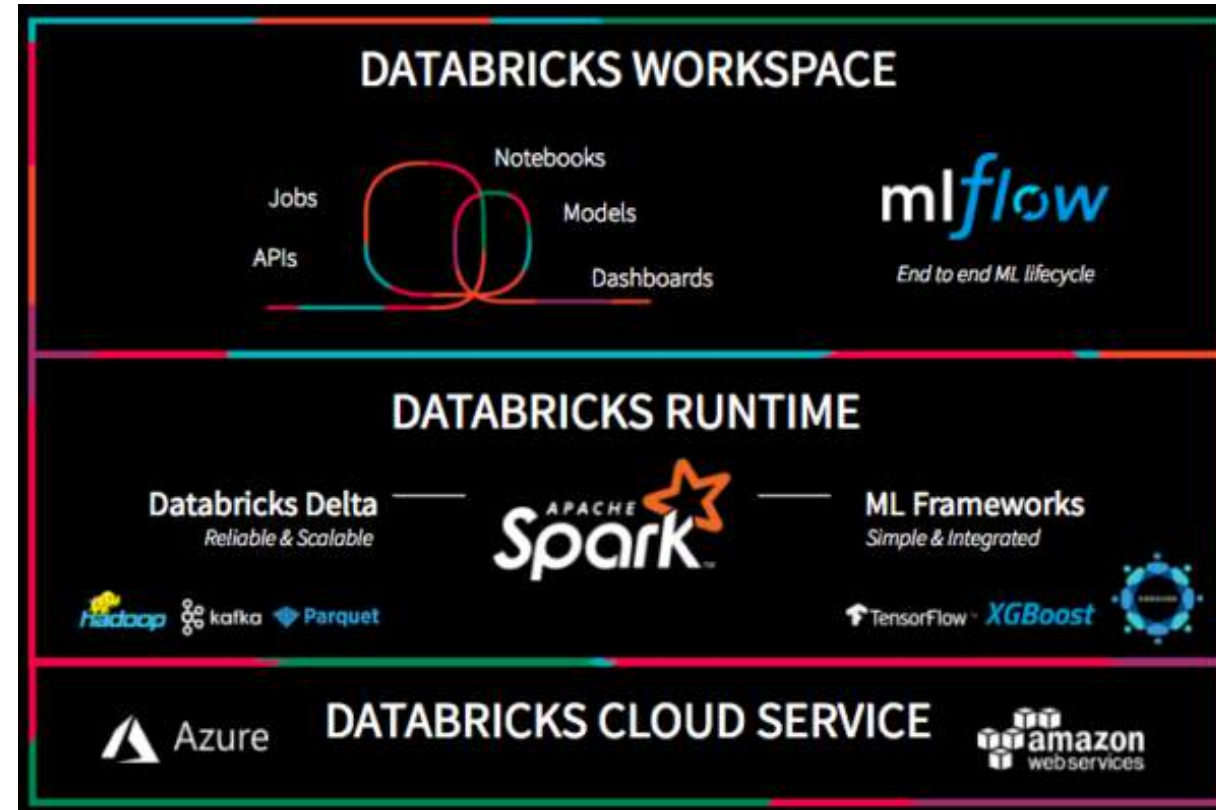
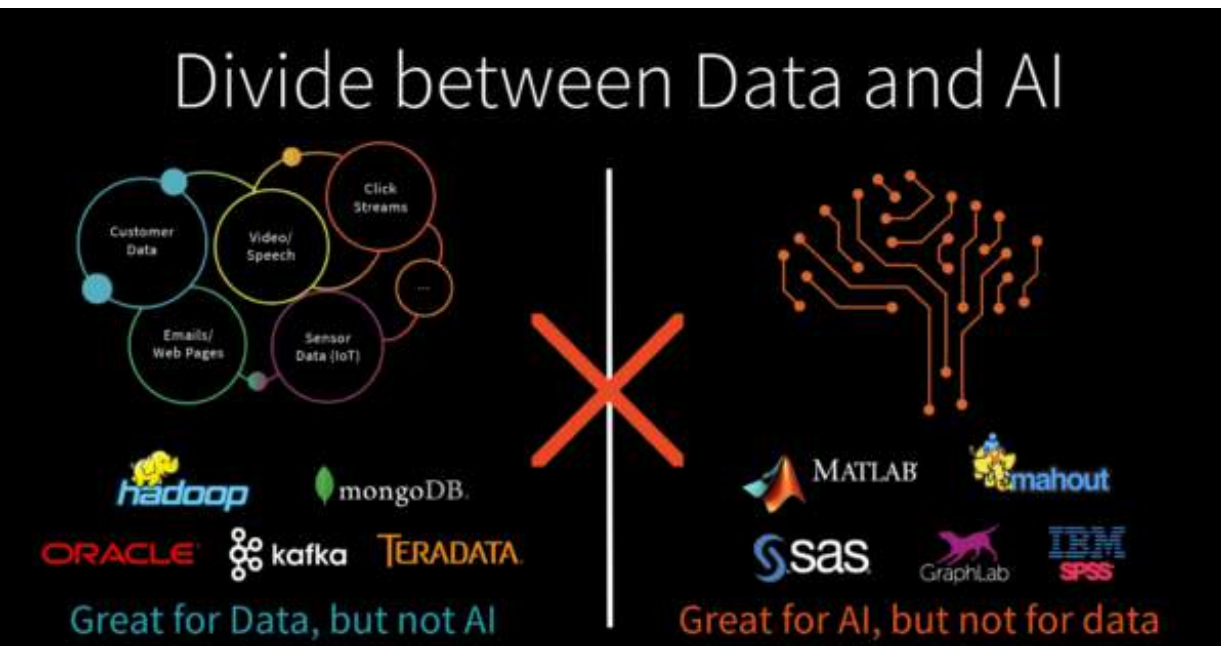
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* source: <https://databricks.com/blog/2018/06/05/distributed-deep-learning-made-simple.html>

BigData + AI: Databricks

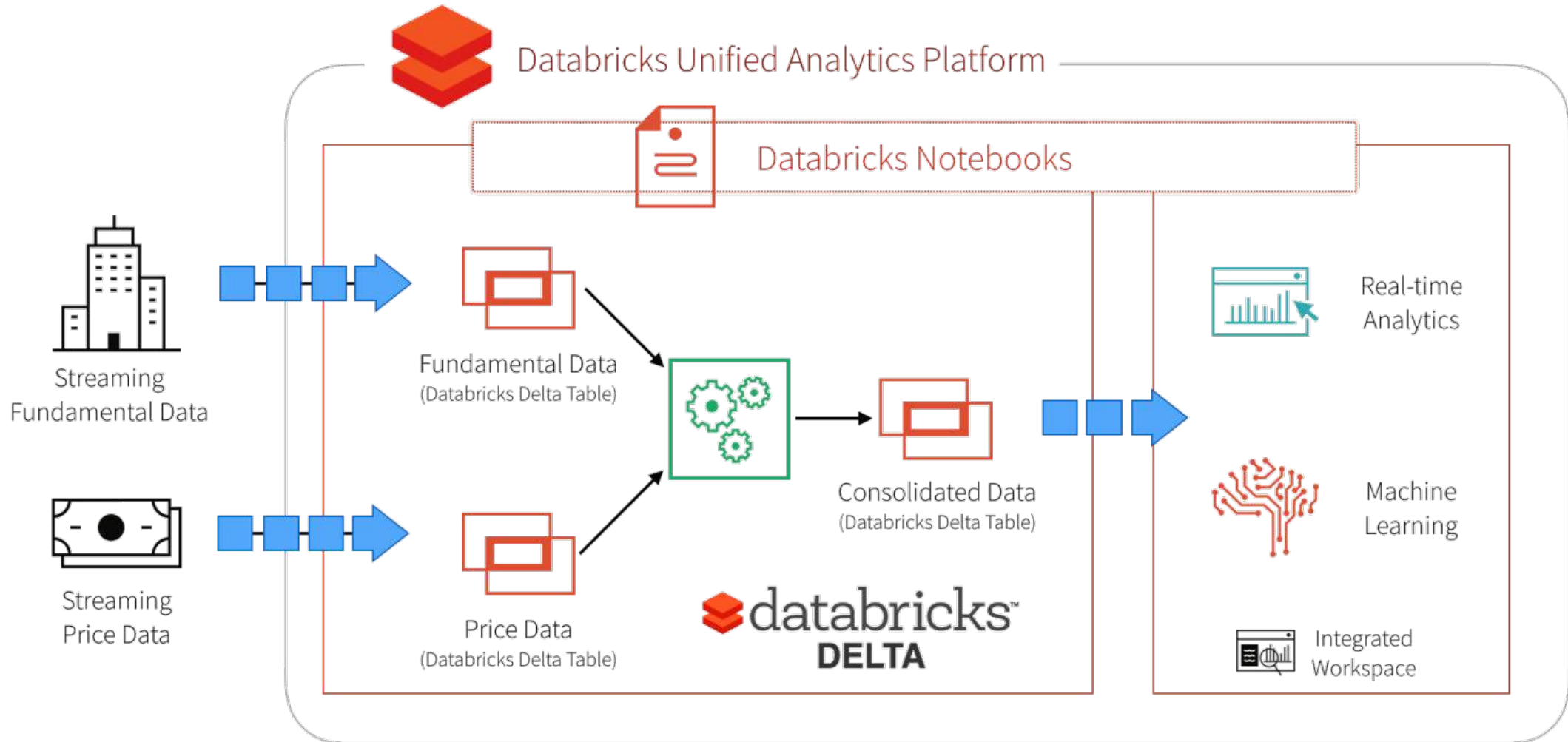
Databricks Unified Analytics Platform



<https://databricks.com/blog/2018/06/05/accelerating-innovation-with-unified-analytics.html>

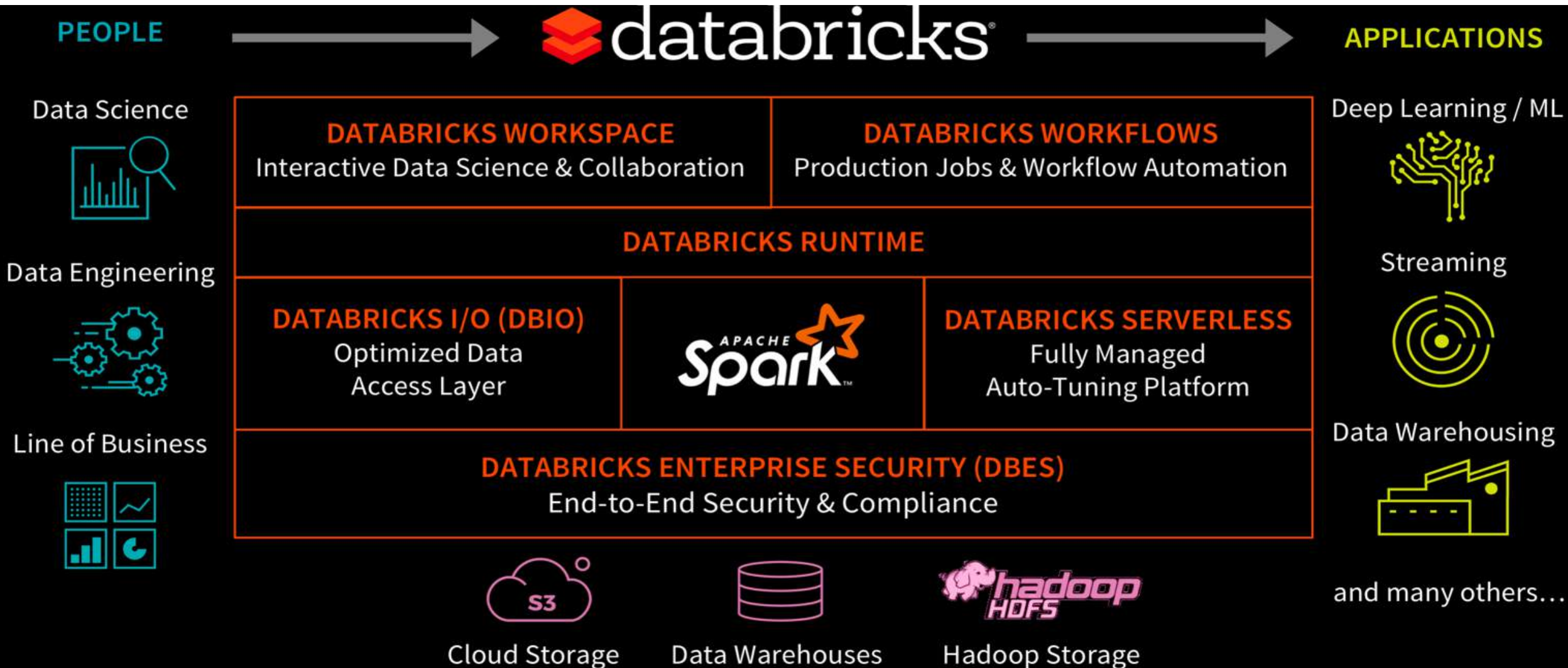
BigData + AI: Databricks

Databricks Unified Analytics Platform

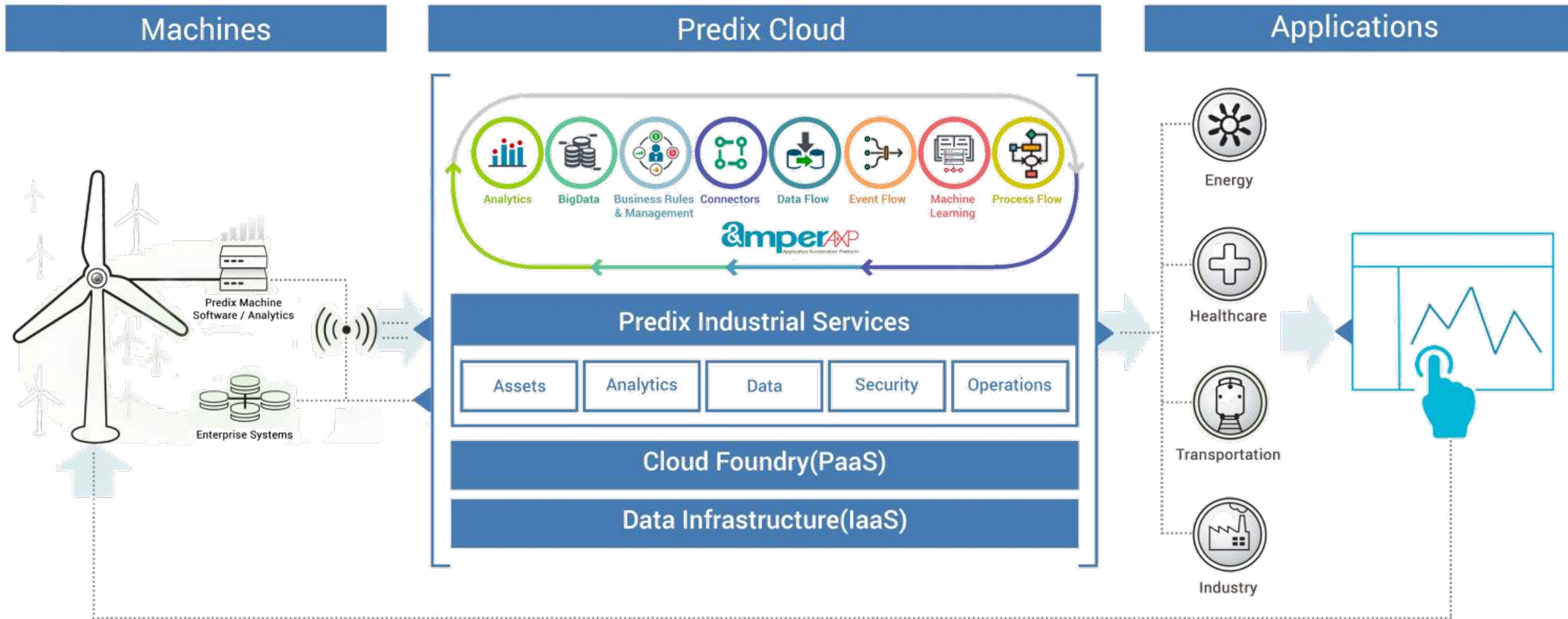


BigData + AI: Databricks

Databricks Unified Analytics Platform



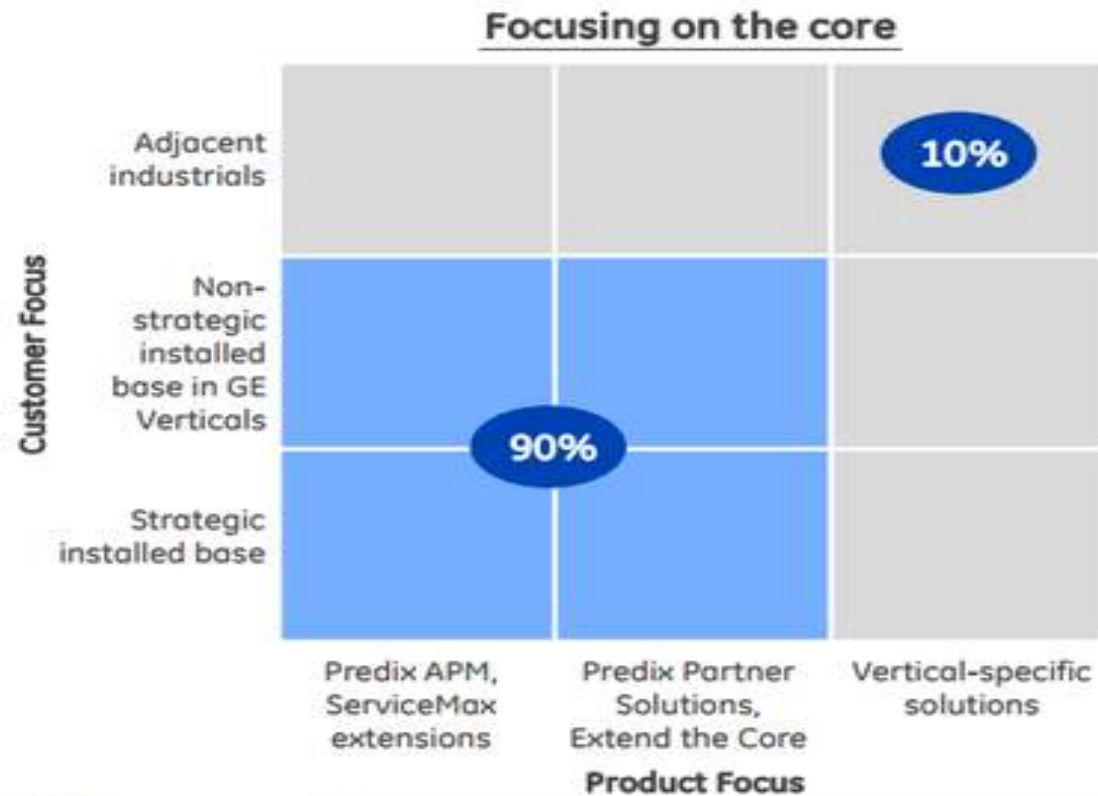
IoT + AI: GE Predix



* source: <http://www.amperxp.com/ge-predix-and-amper/>

IoT + AI: GE Predix

4 Digital



- Approach
- 1 Lead with Predix applications that drive customer outcomes: APM, OPM, and ServiceMax
 - 2 Focus spend on Predix platform differentiation: asset model, Edge to Cloud, Digital Twin
 - 3 Partner for technology that is not differentiated (i.e. Cloud)
 - 4 Prioritize go to market around GE business verticals where win rate is ~2x higher

Customer examples



- APM used at 1.3 GW power plant
- 1% efficiency gain on mixed fleet
- ~\$18MM annual customer value
- APM used to improve asset availability
- \$1.4MM saved in production losses
- ~\$1.3MM revenue increase



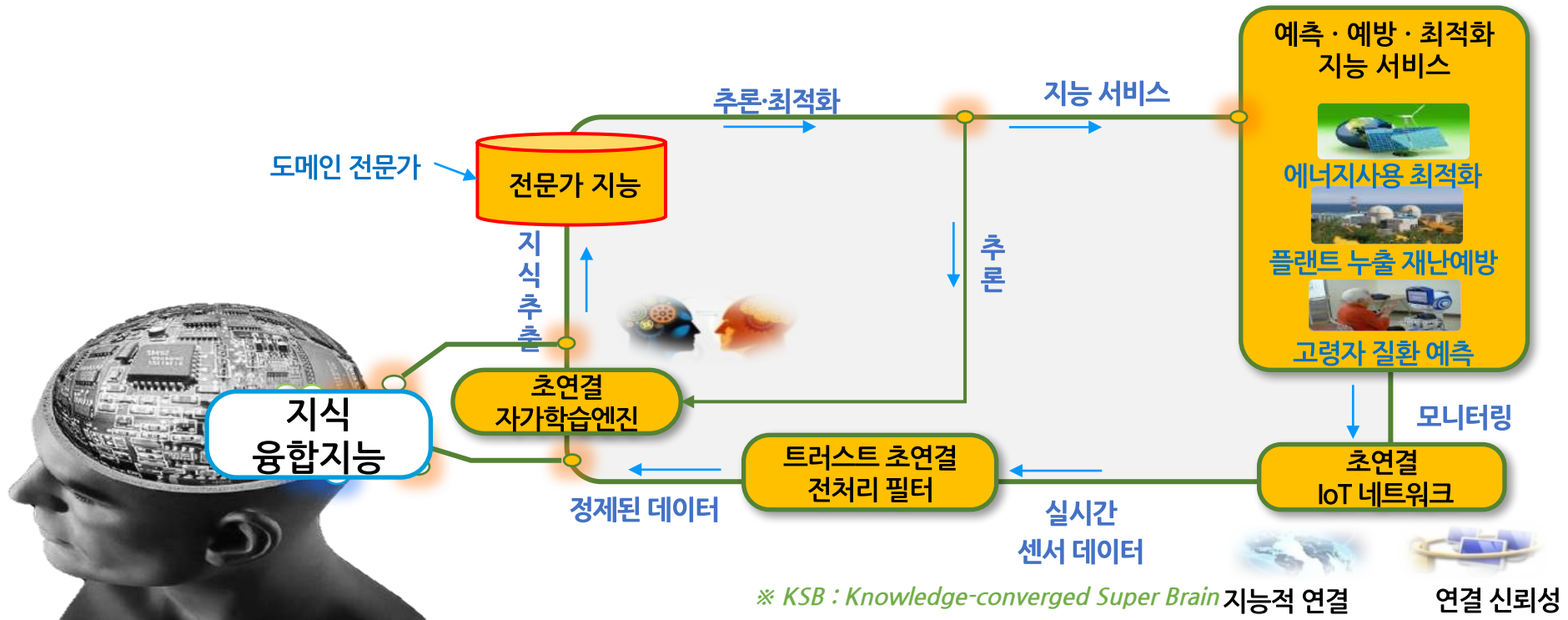
Targeting \$1B+ Predix-powered revenue and \$0.4B of cost out in 2018

* source: <https://www.zdnet.com/article/ge-to-hone-digital-efforts-leverage-additive-manufacturing-as-it-focuses-on-core-businesses/>

III. Strategies of KSB Framework

Knowledge-converged Super Brain

- ① 상태를 모니터링하여 데이터를 수집 ② 수집된 데이터를 정제 ③ 기계학습·지식추출하고
- ④ 도메인의 전문가지식과 융합하여 추론·최적화함으로써 ⑤ 예측·예방·최적화 지능 서비스 제공

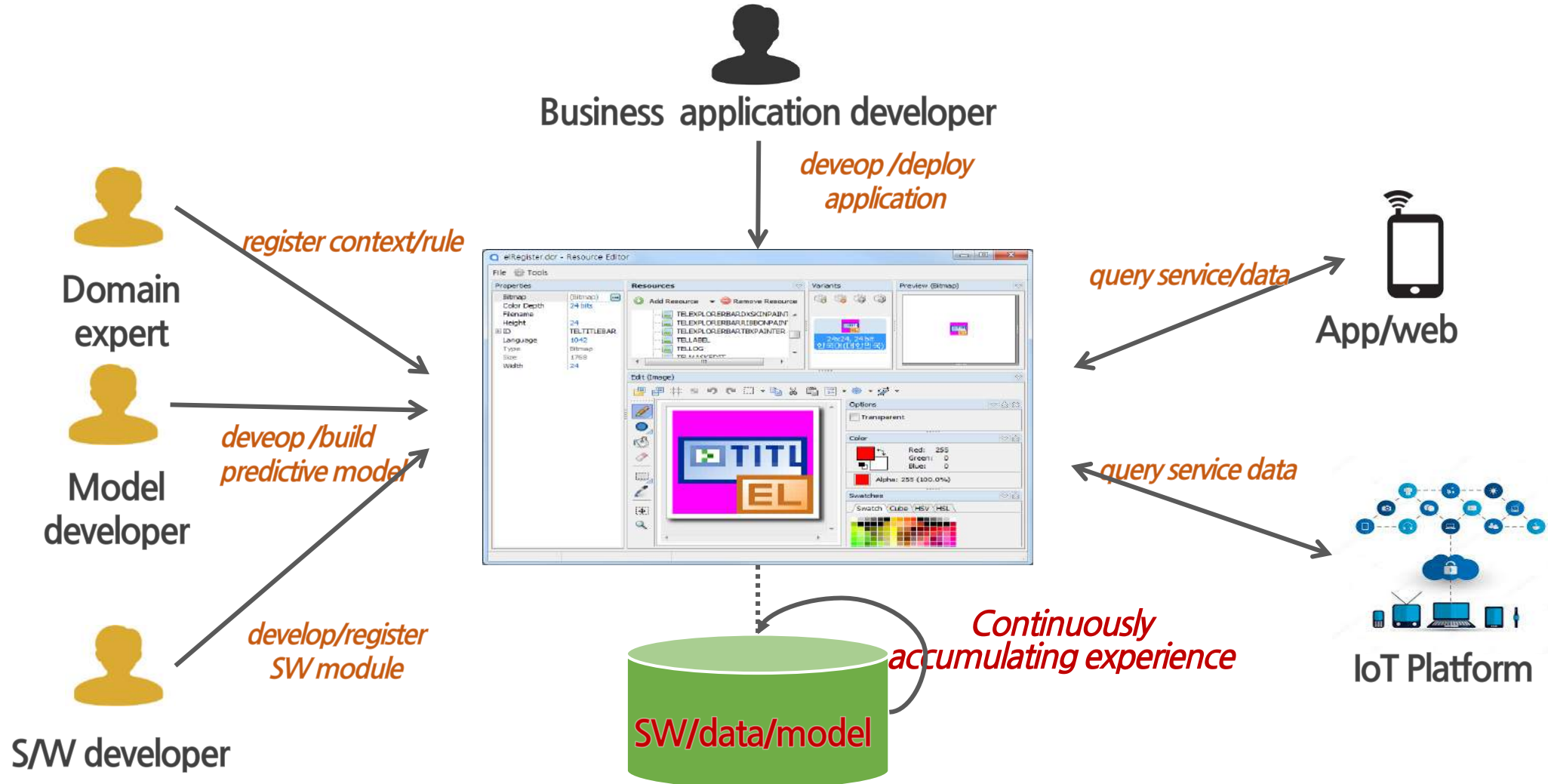


Knowledge-converged Super Brain

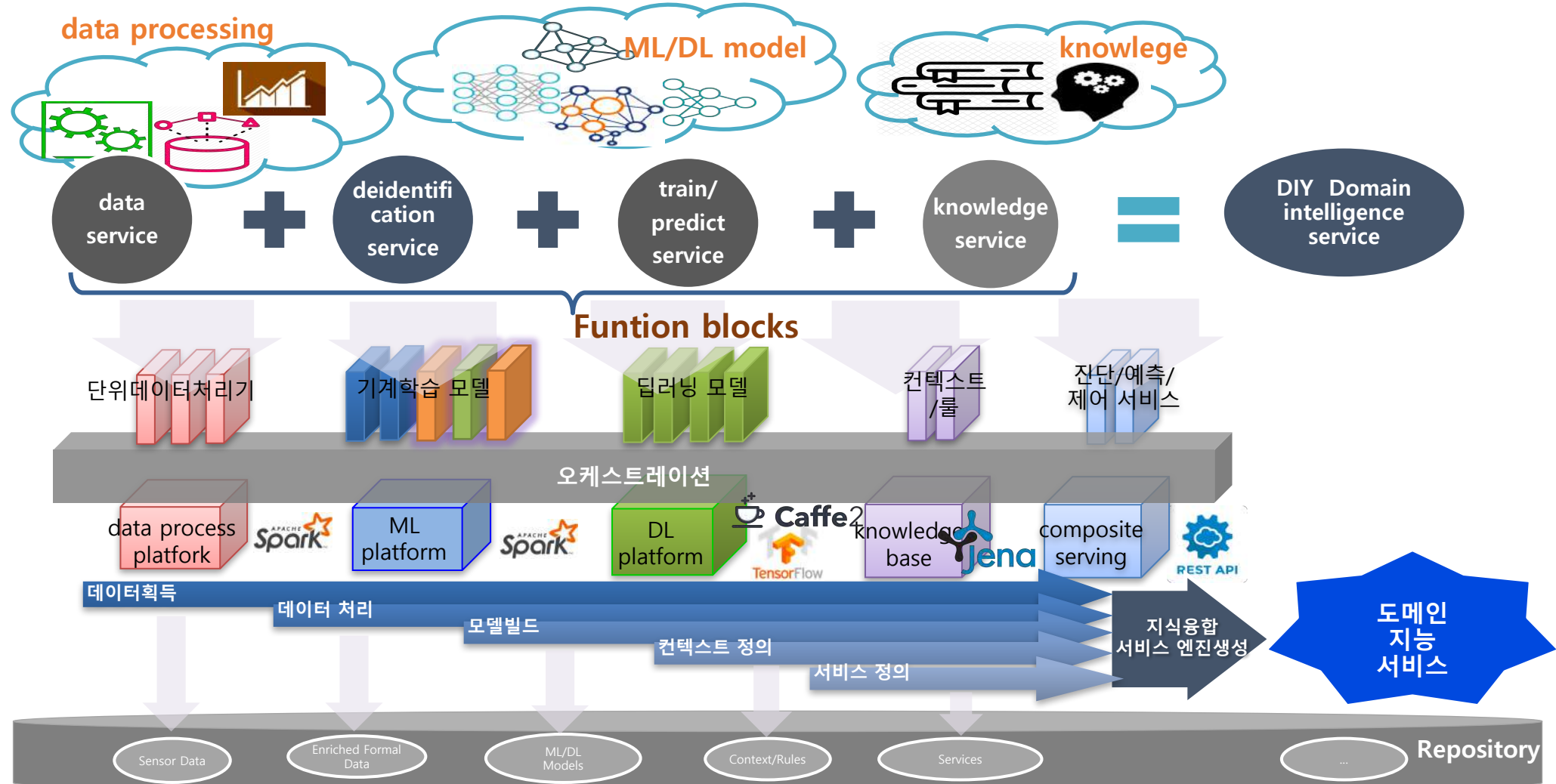


Energy/Plant/Health

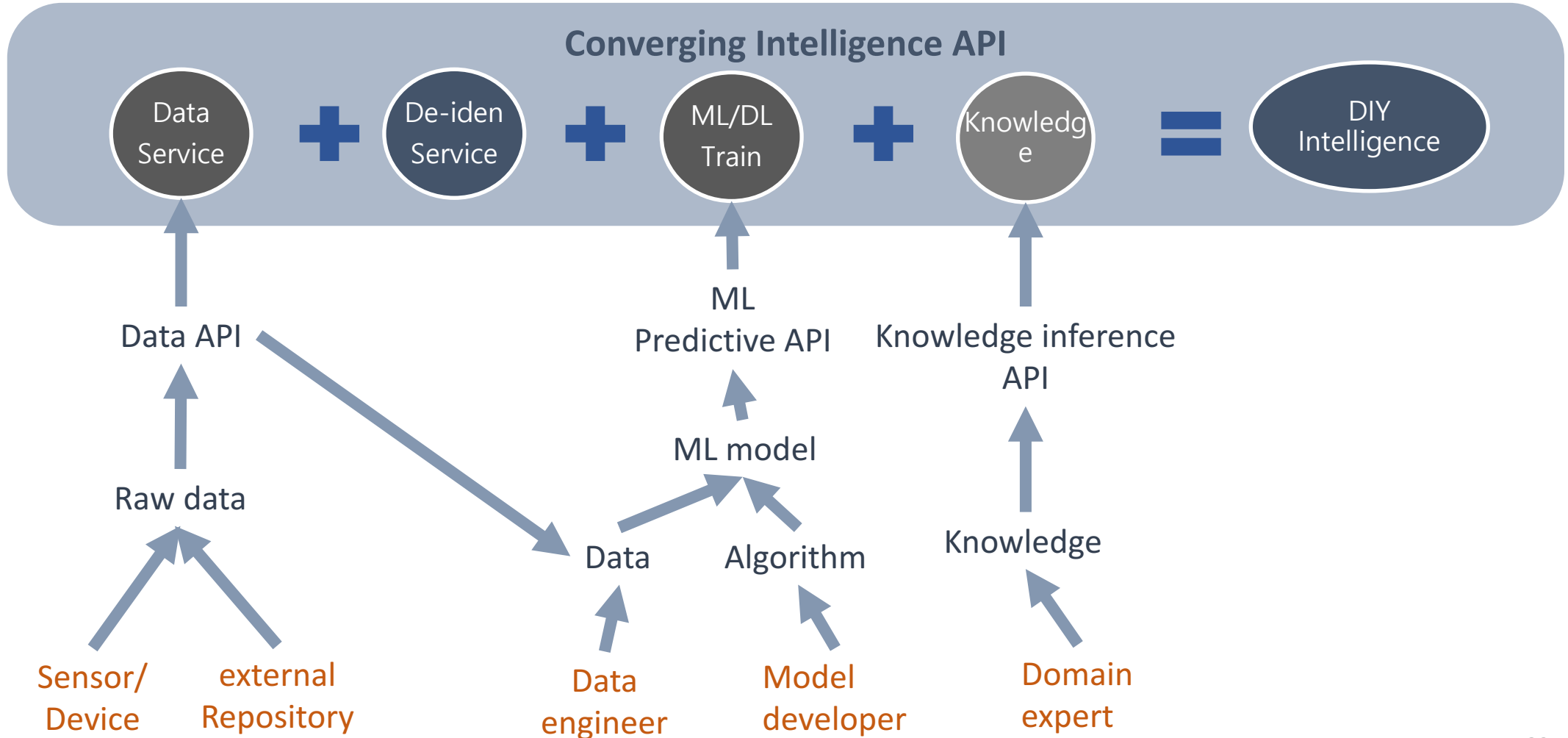
KSB Framework Approach



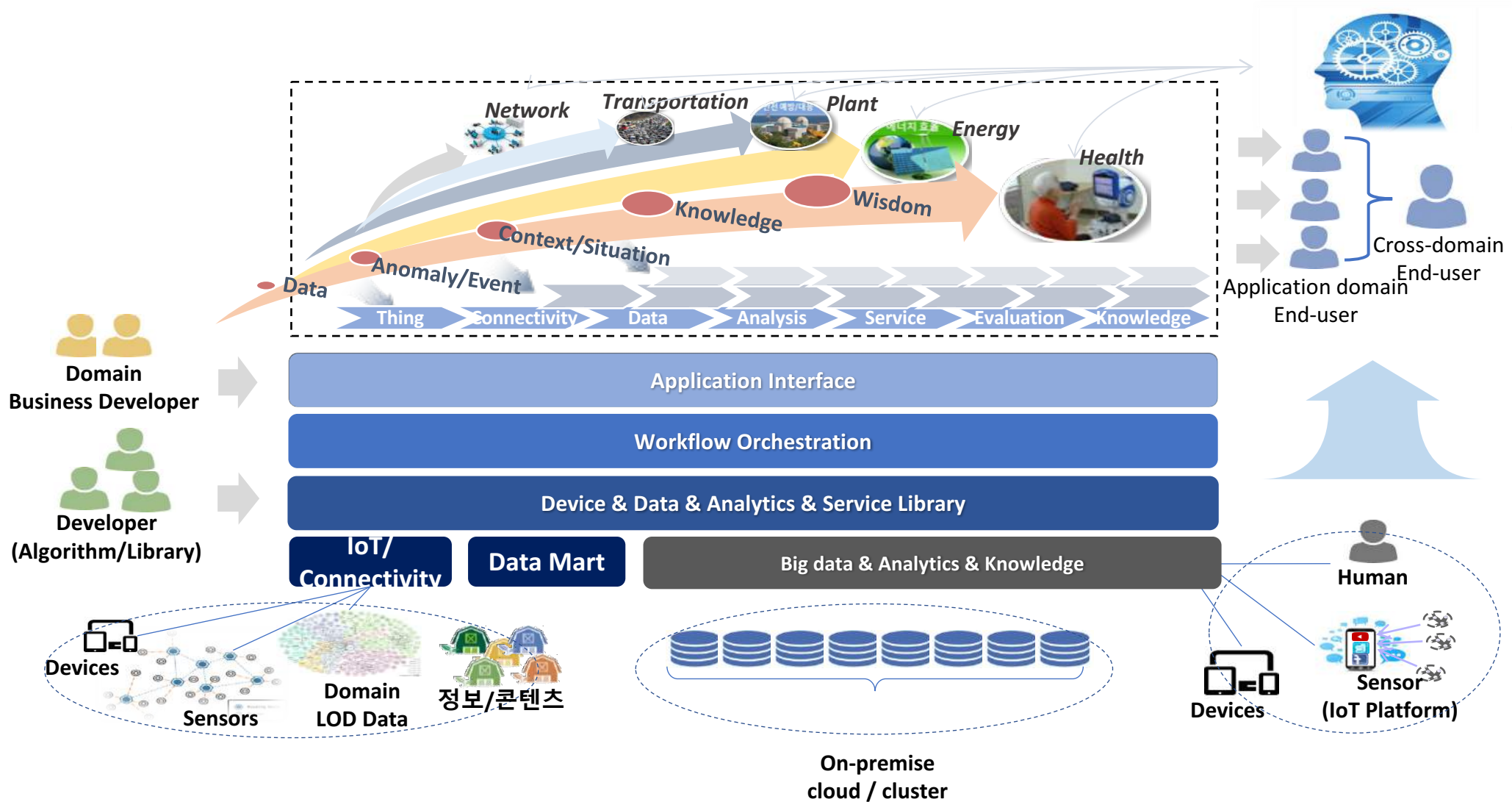
KSB Framework Concept



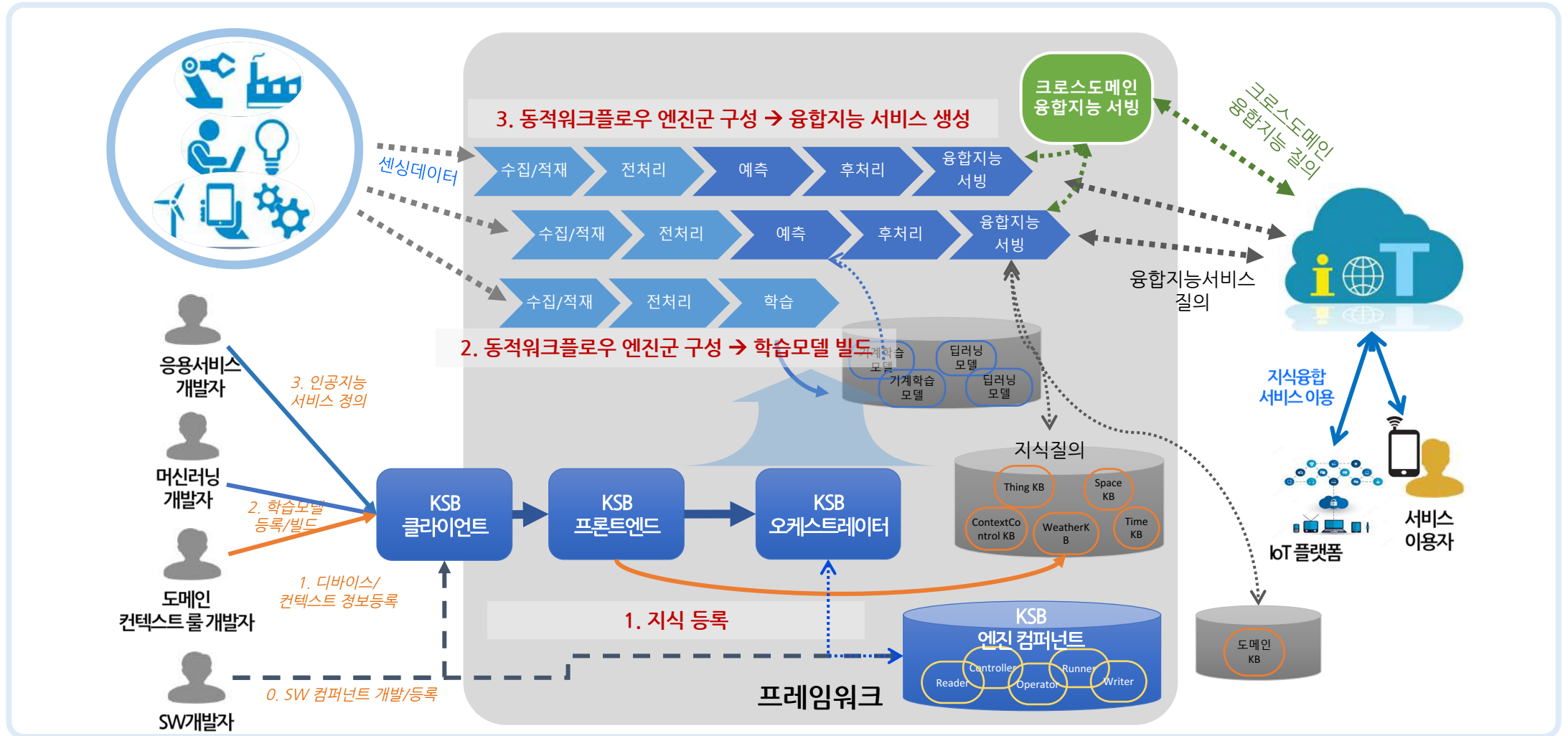
KSB Framework Concept



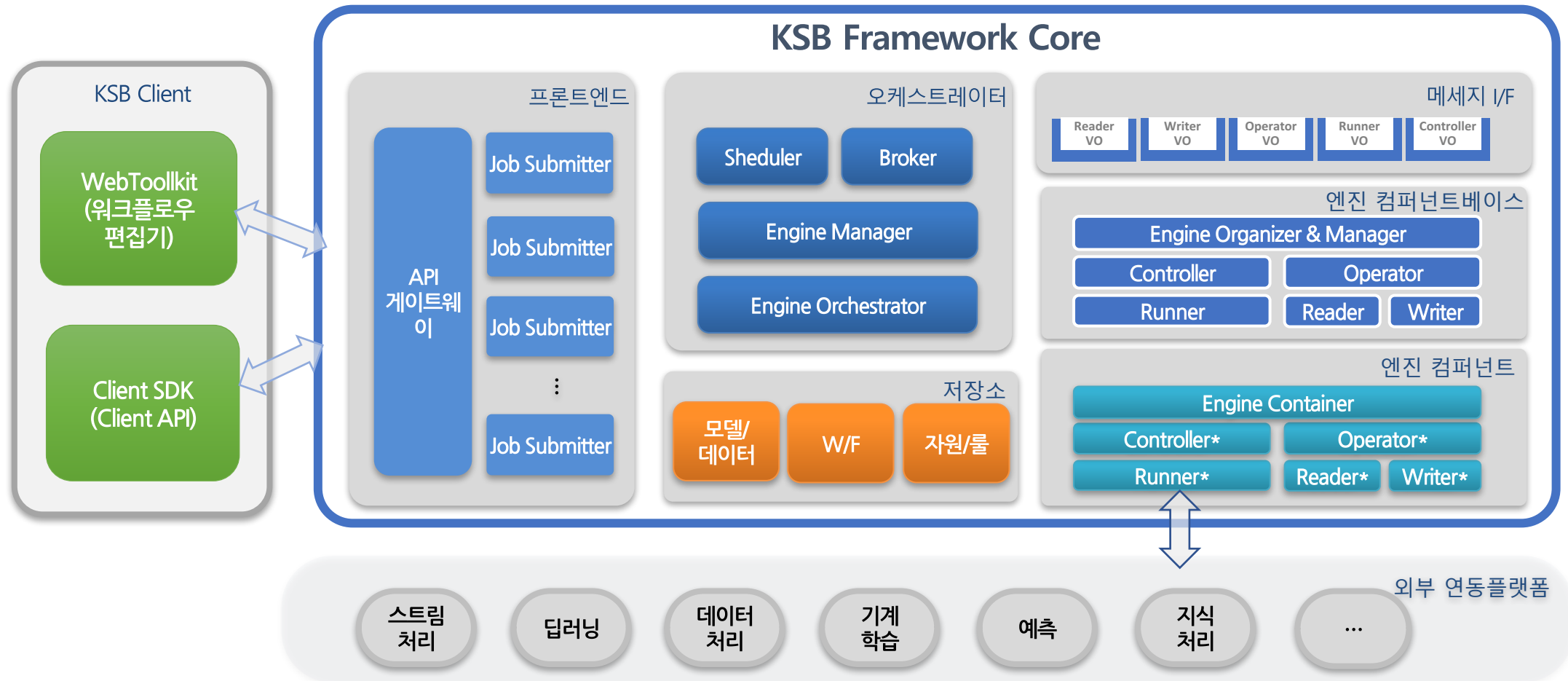
KSB Framework Overview



KSB Framework Overview



KSB Framework Architecture



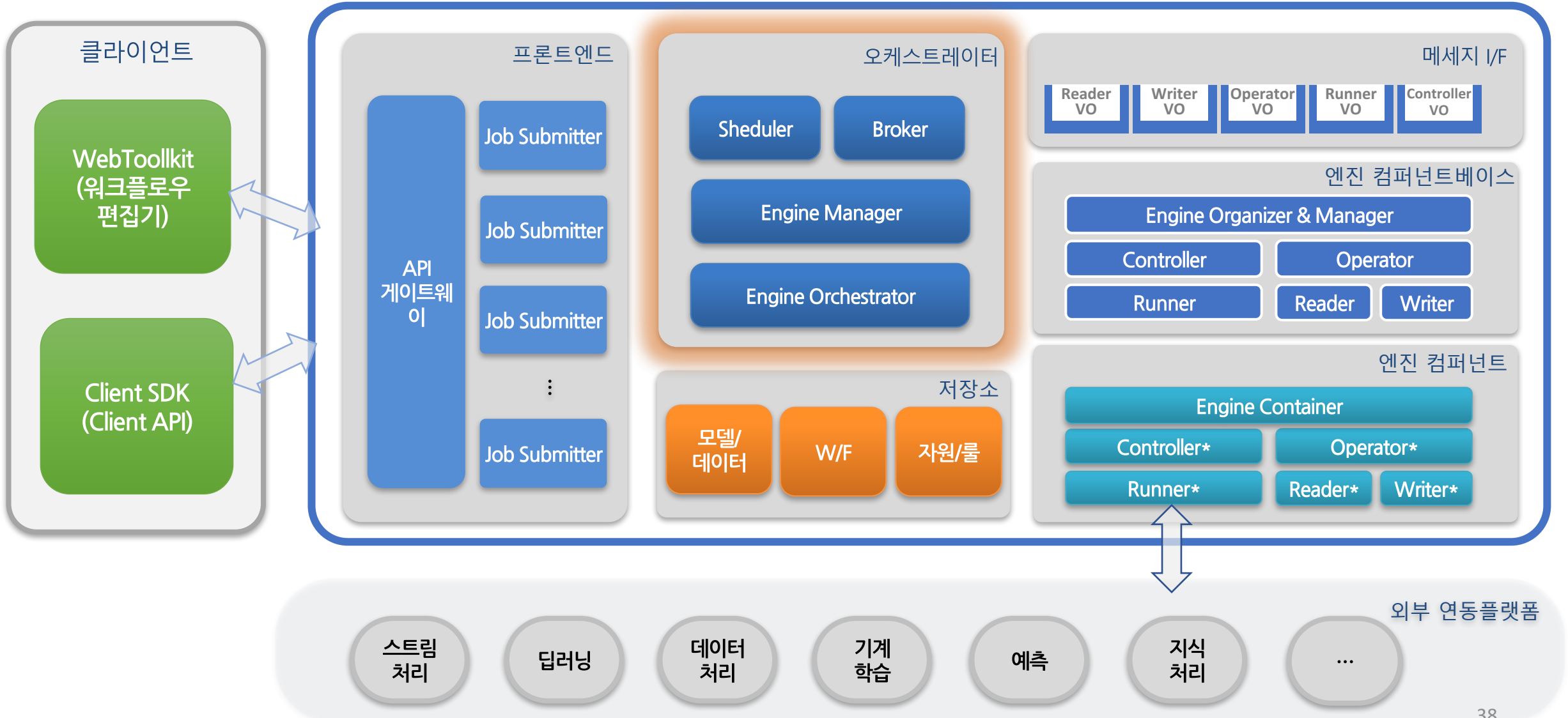
KSB 인공지능 프레임워크 Core



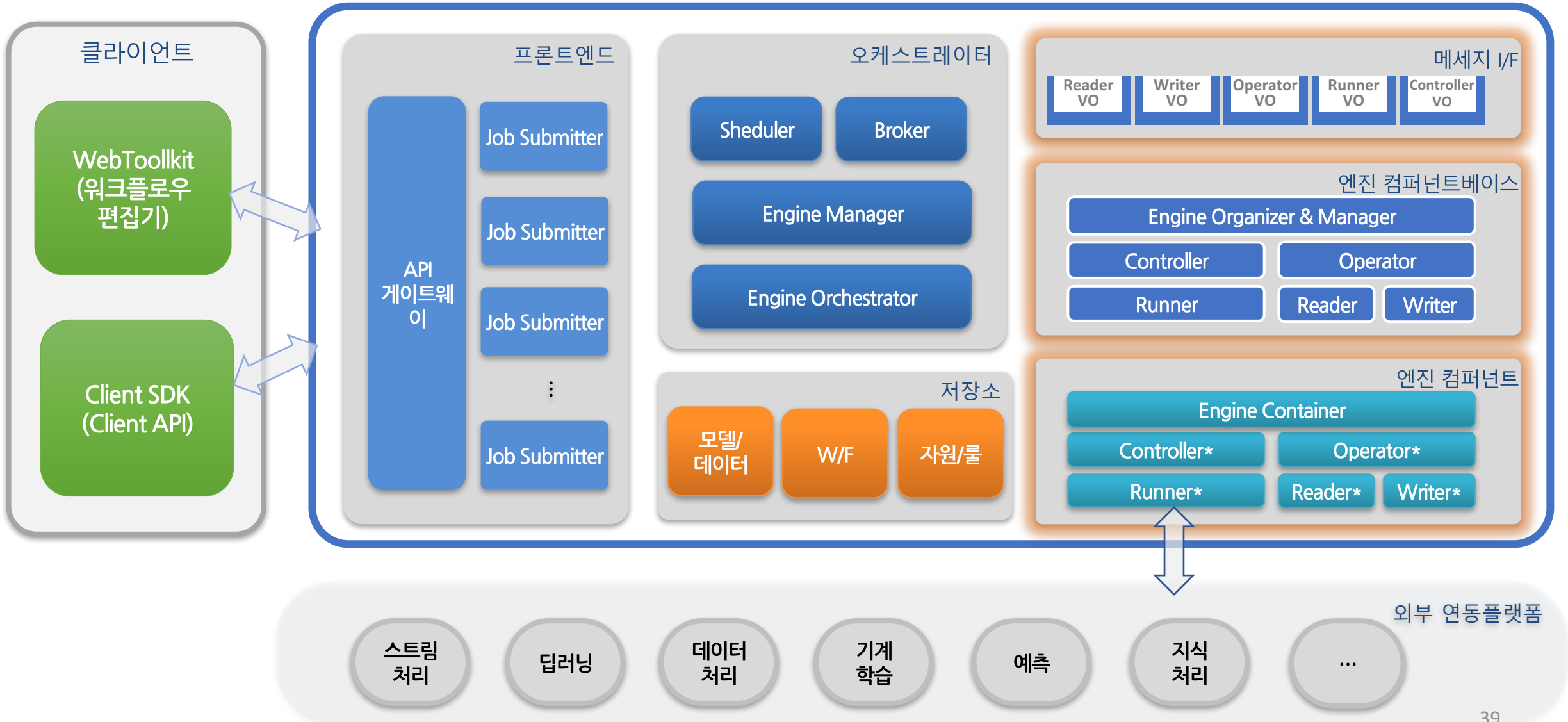
KSB 인공지능 프레임워크 Core



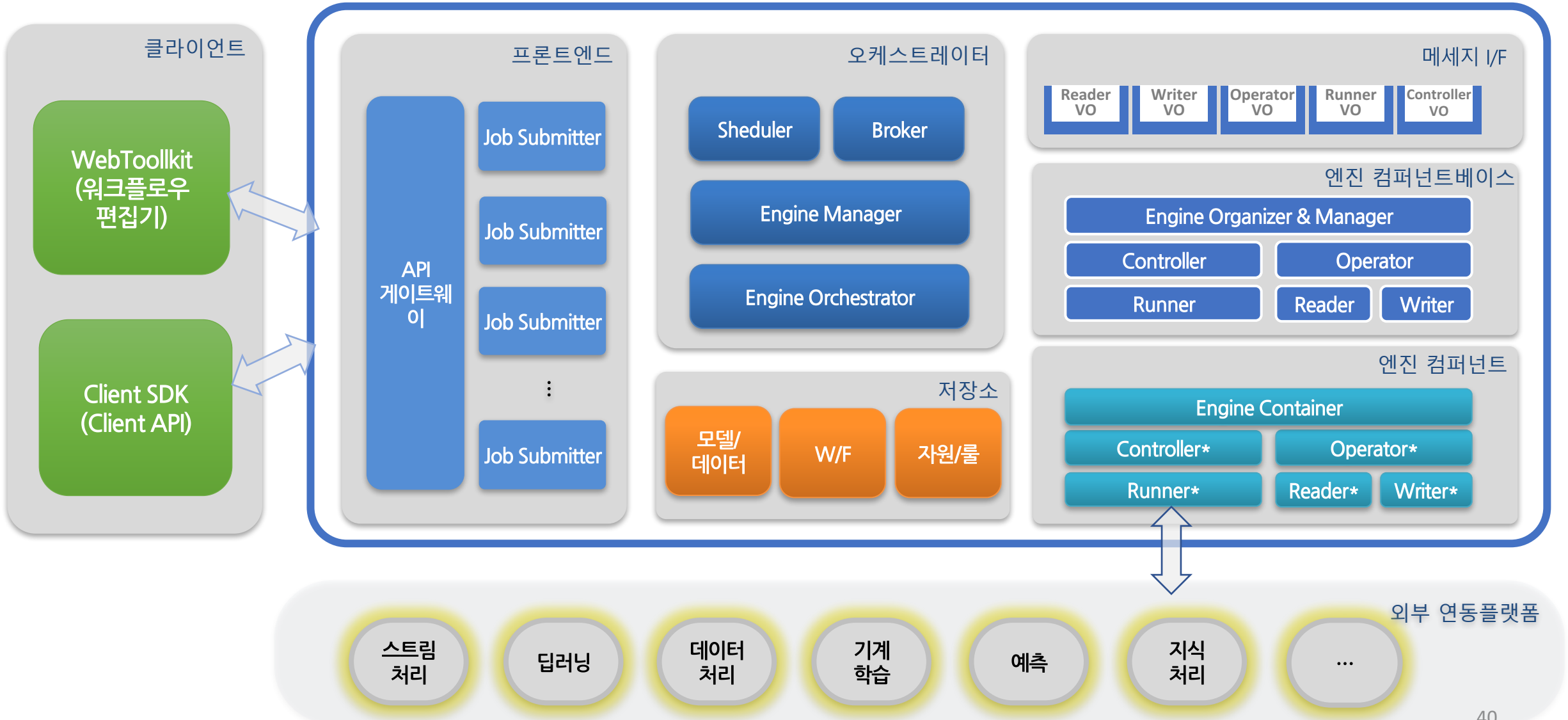
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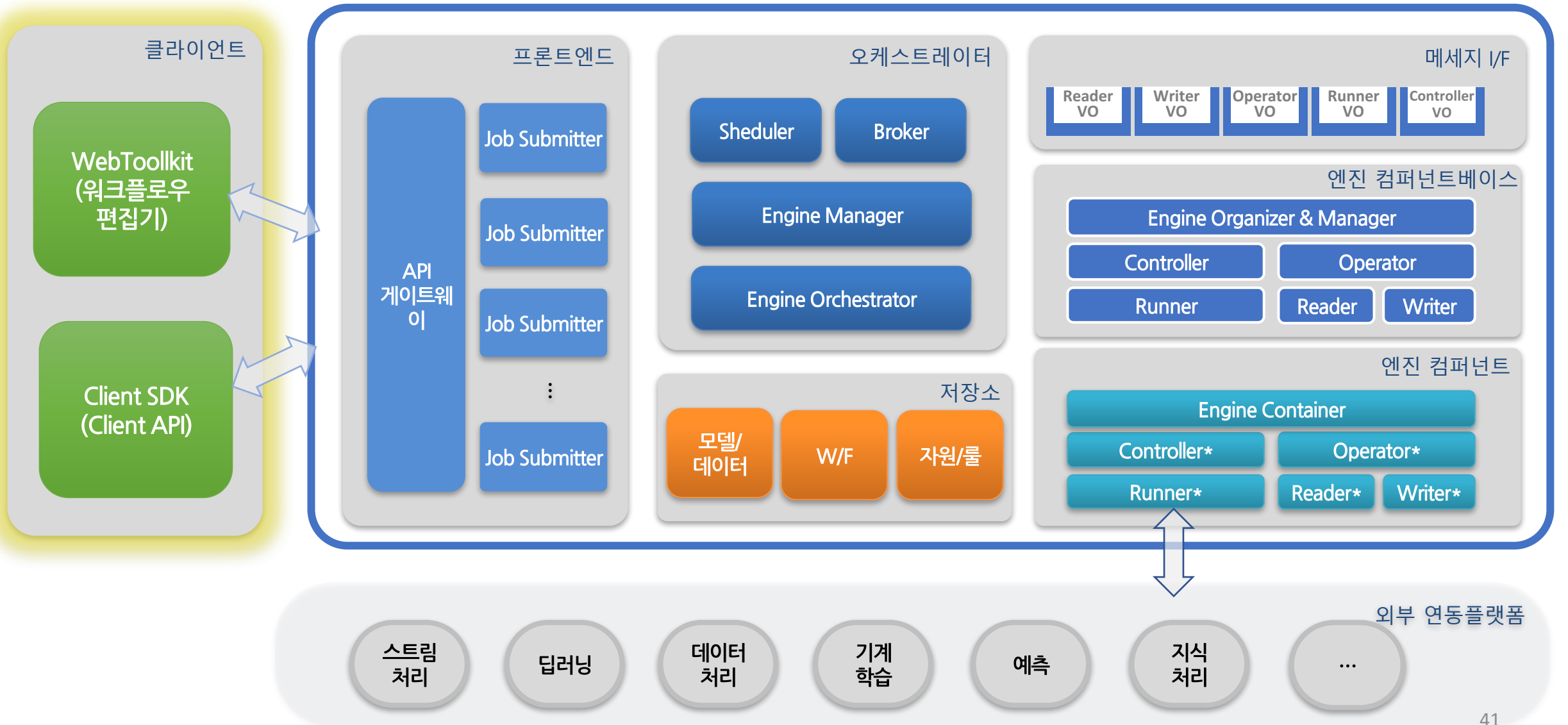
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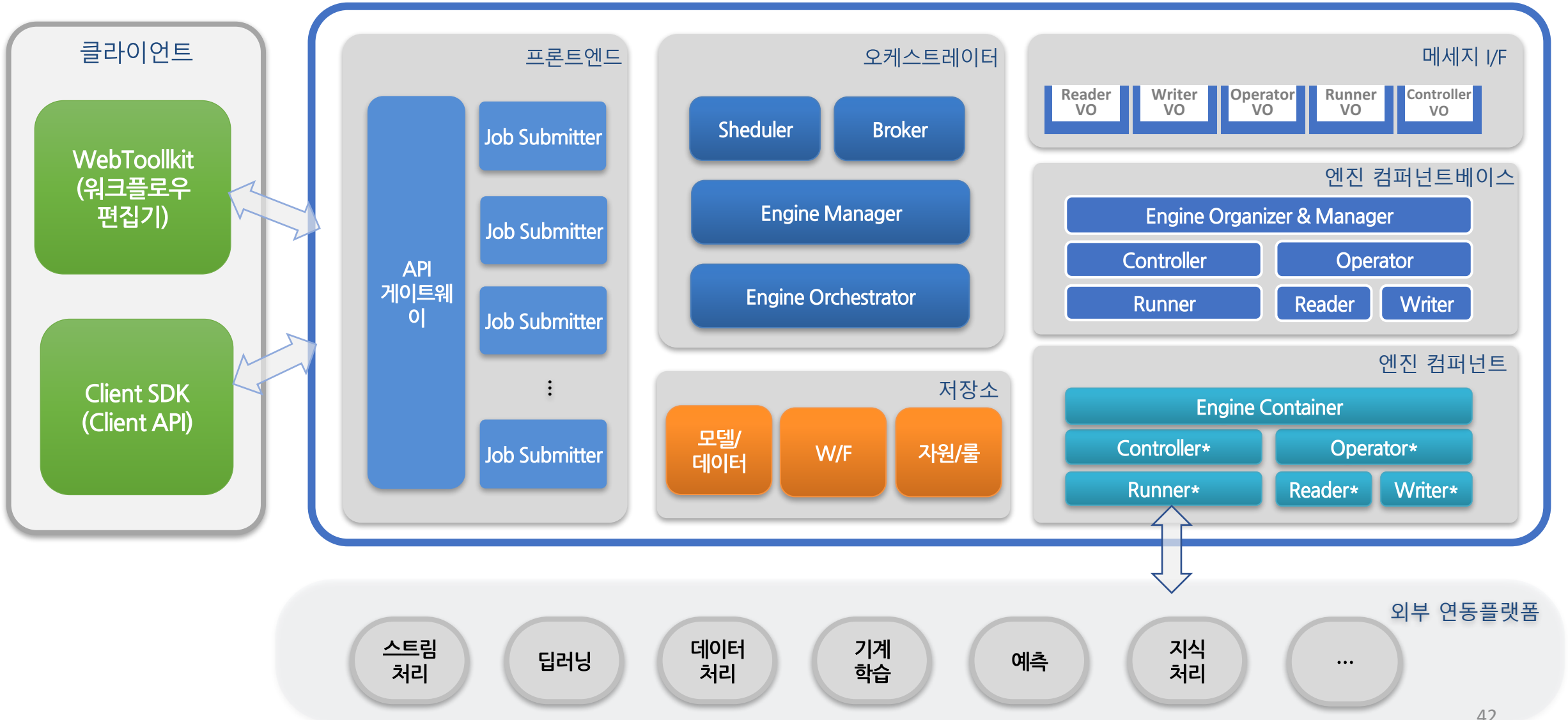
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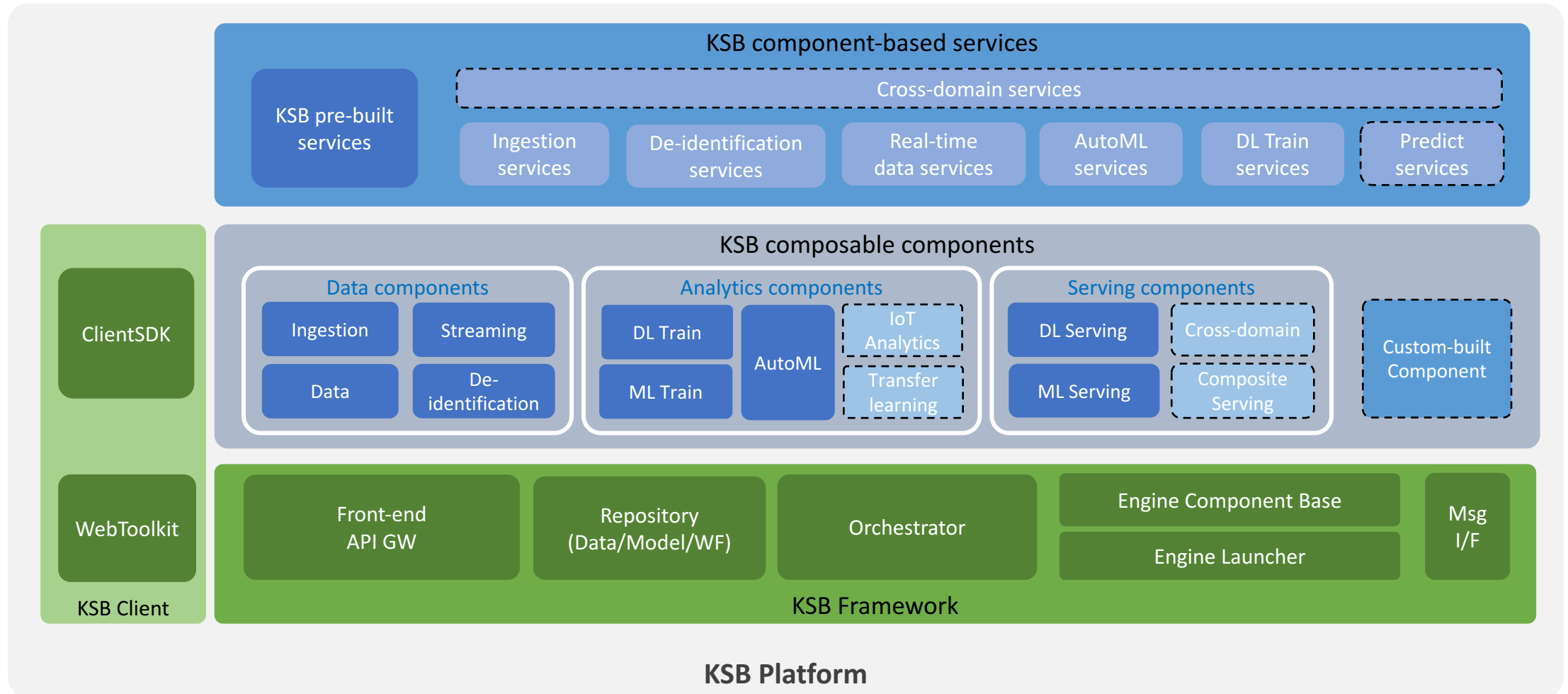
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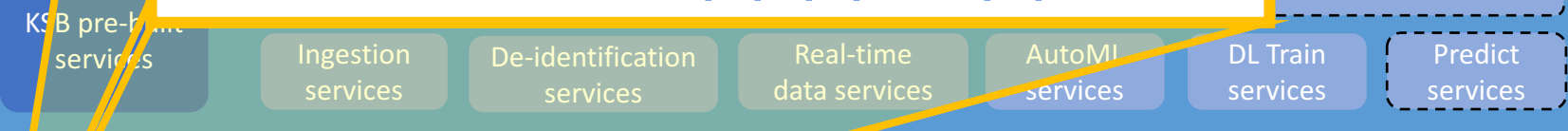


KSB Framework Stack

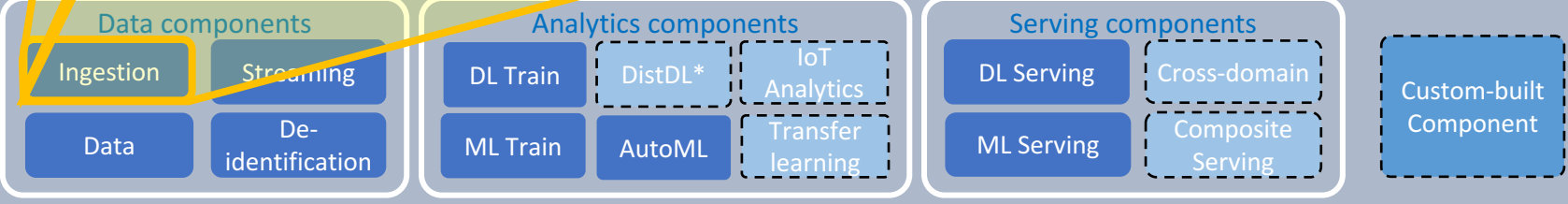




IoE 스트림 데이터 수집/적재



KSB composable components



KSB Client



KSB Framework Core

KSB Platform



KSB pre-built services

Ingestion services

De-identification services

Real-time data services

AutoML services

DL Train services

Predict services

KSB composable components

Data components

Ingestion

Streaming

Data

De-identification

Analytics components

DL Train

DistDL*

IoT Analytics

ML Train

AutoML

Transfer learning

Serving components

DL Serving

Cross-domain

ML Serving

Composite Serving

Custom-built Component

ClientSDK

WebToolkit

KSB Client

Front-end API GW

Repository (Data/Model/WF)

Registry (SW/Service)

Orchestrator

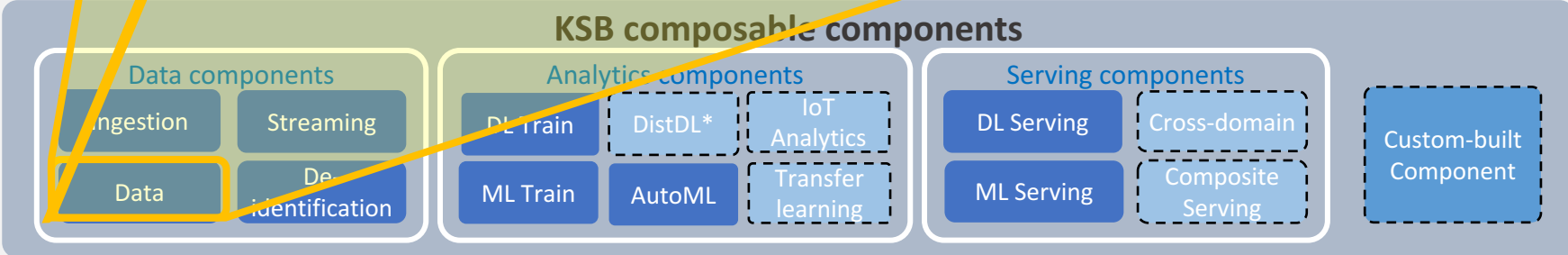
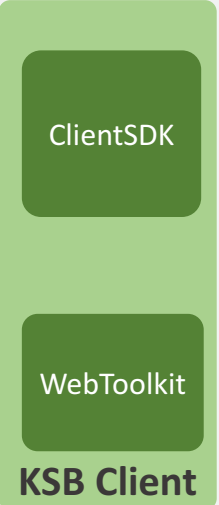
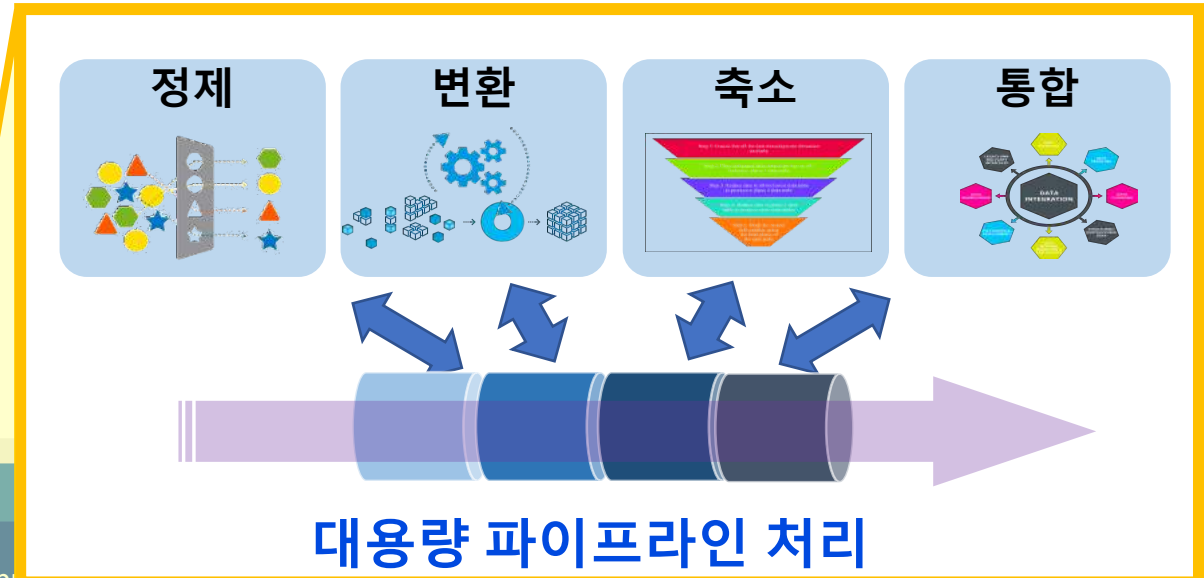
Engine Component Base

Engine Launcher

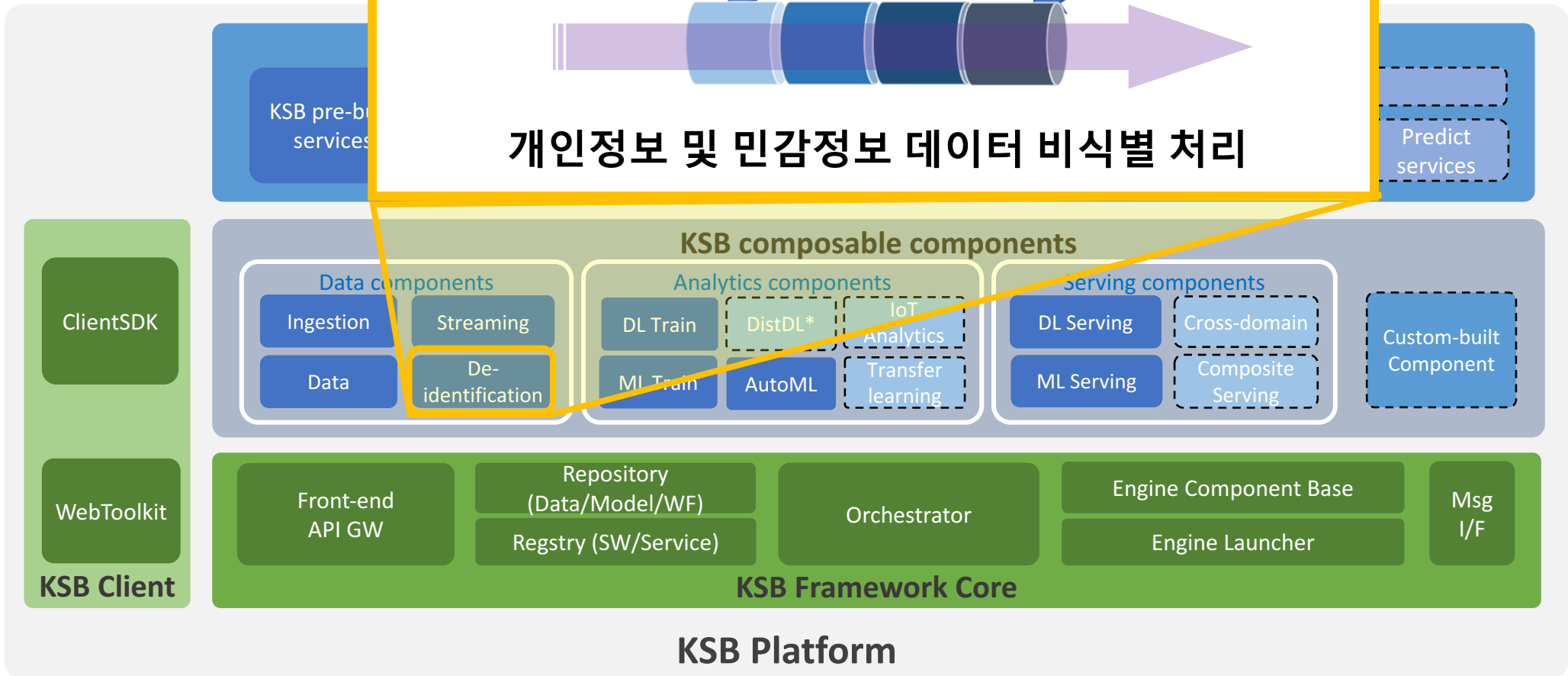
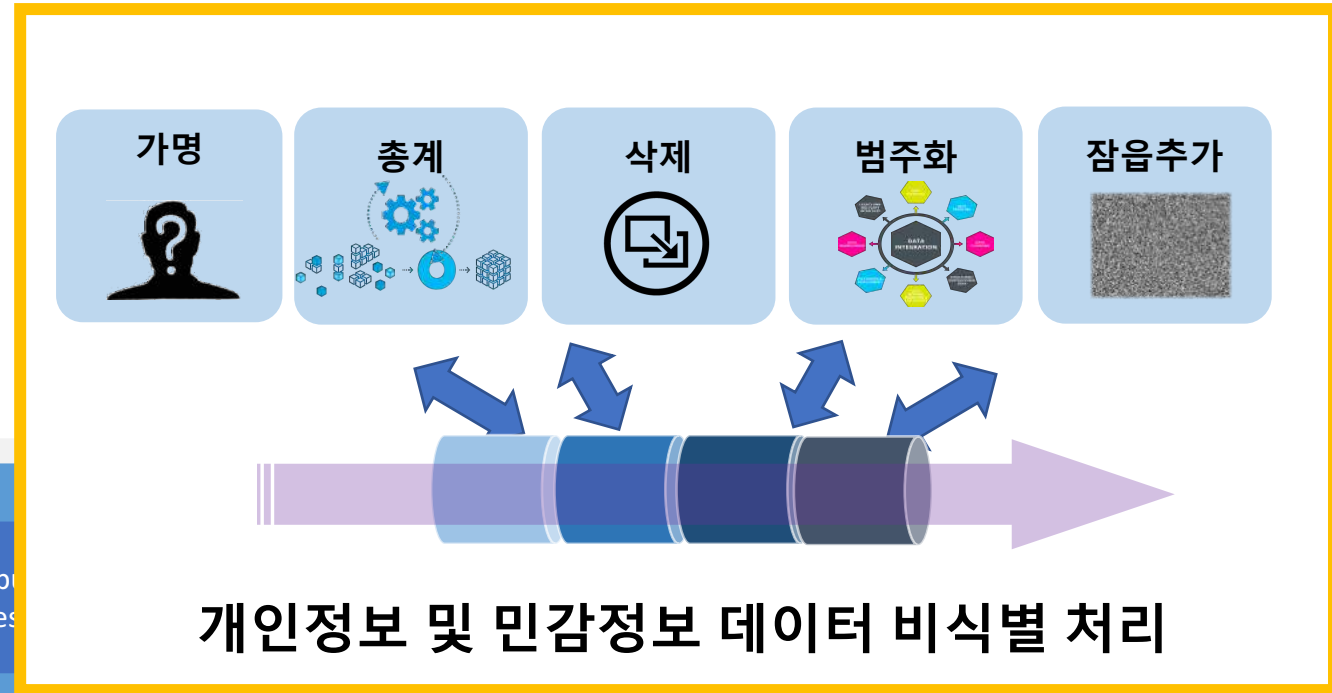
Msg I/F

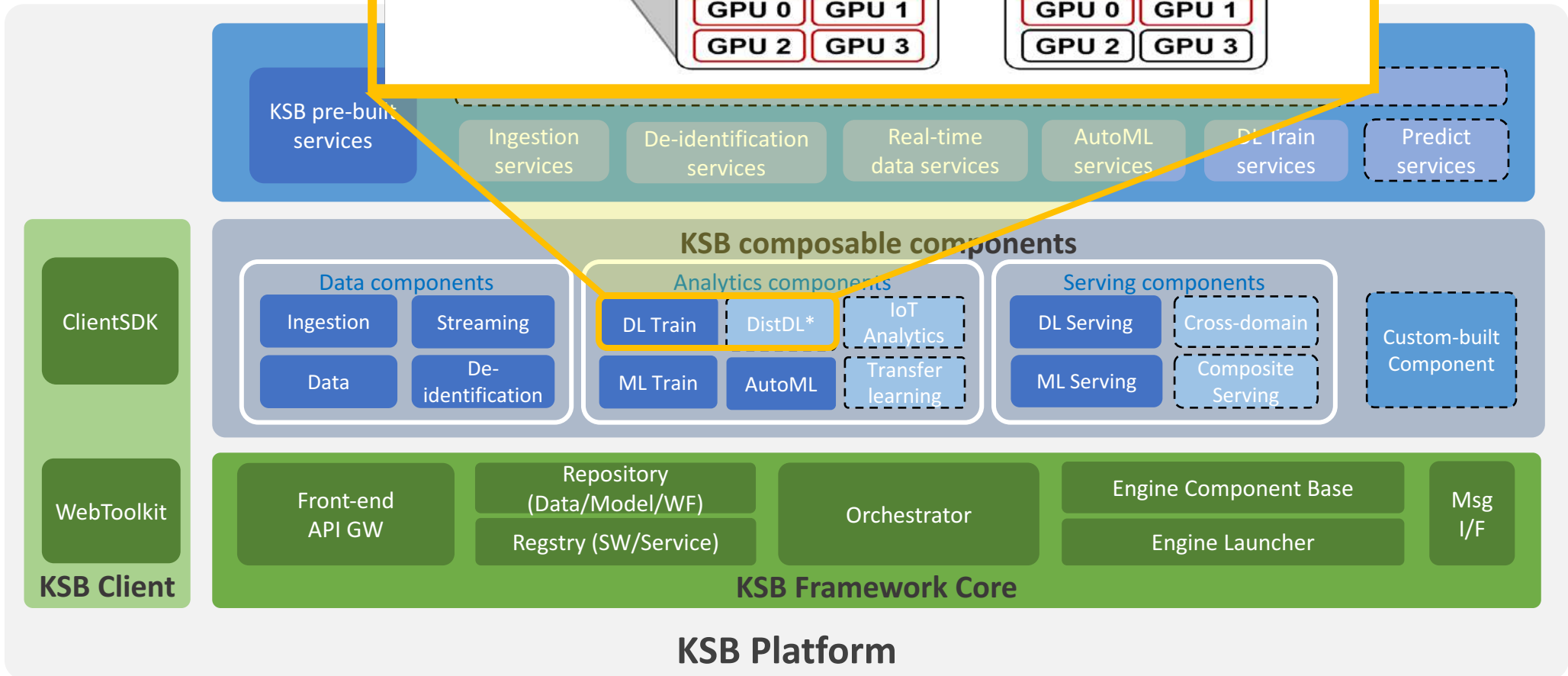
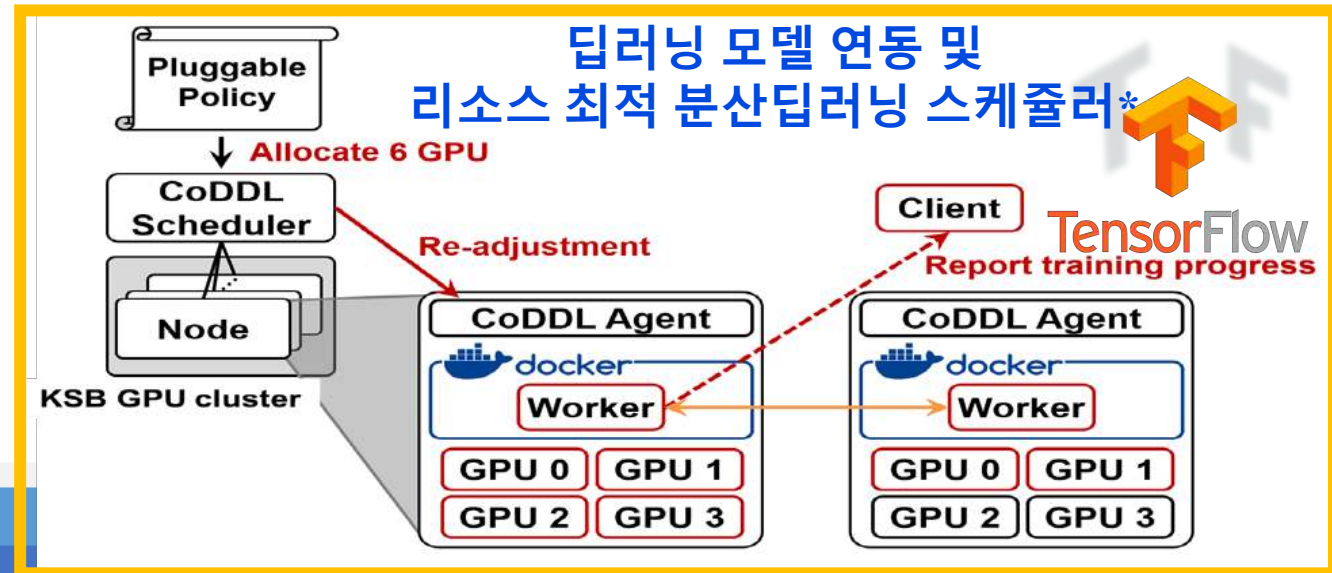
KSB Framework Core

KSB Platform

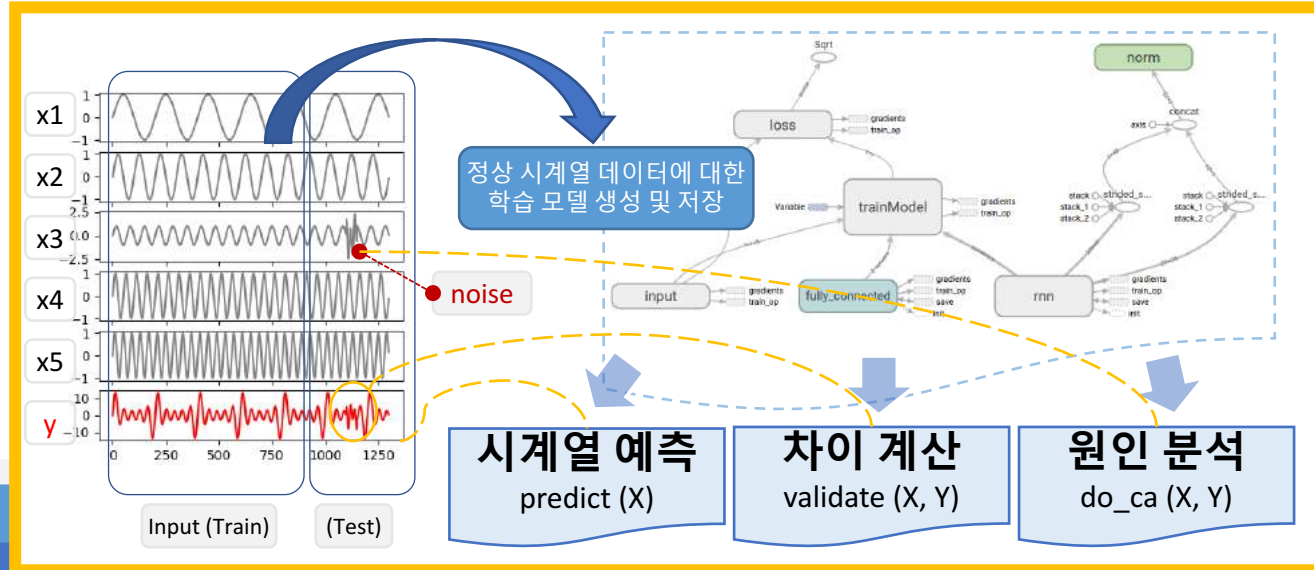


KSB Platform





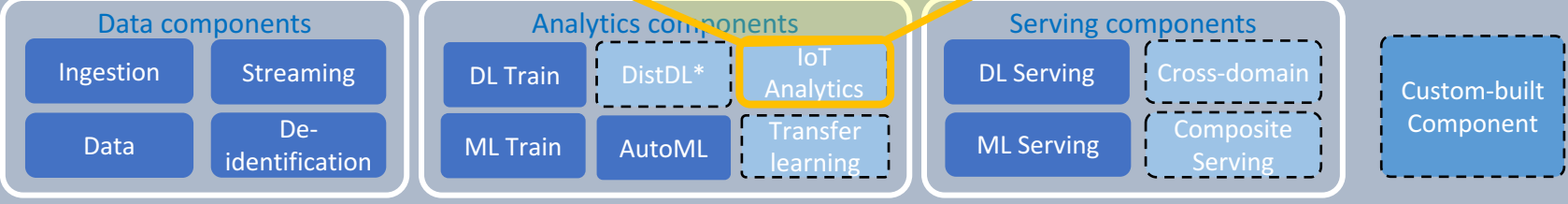
이상감지



KSB pre-built services

- Ingestion services
- De-identification services
- Real-time data services
- AutoML services
- DL Train services
- Predict services

KSB composable components



ClientSDK

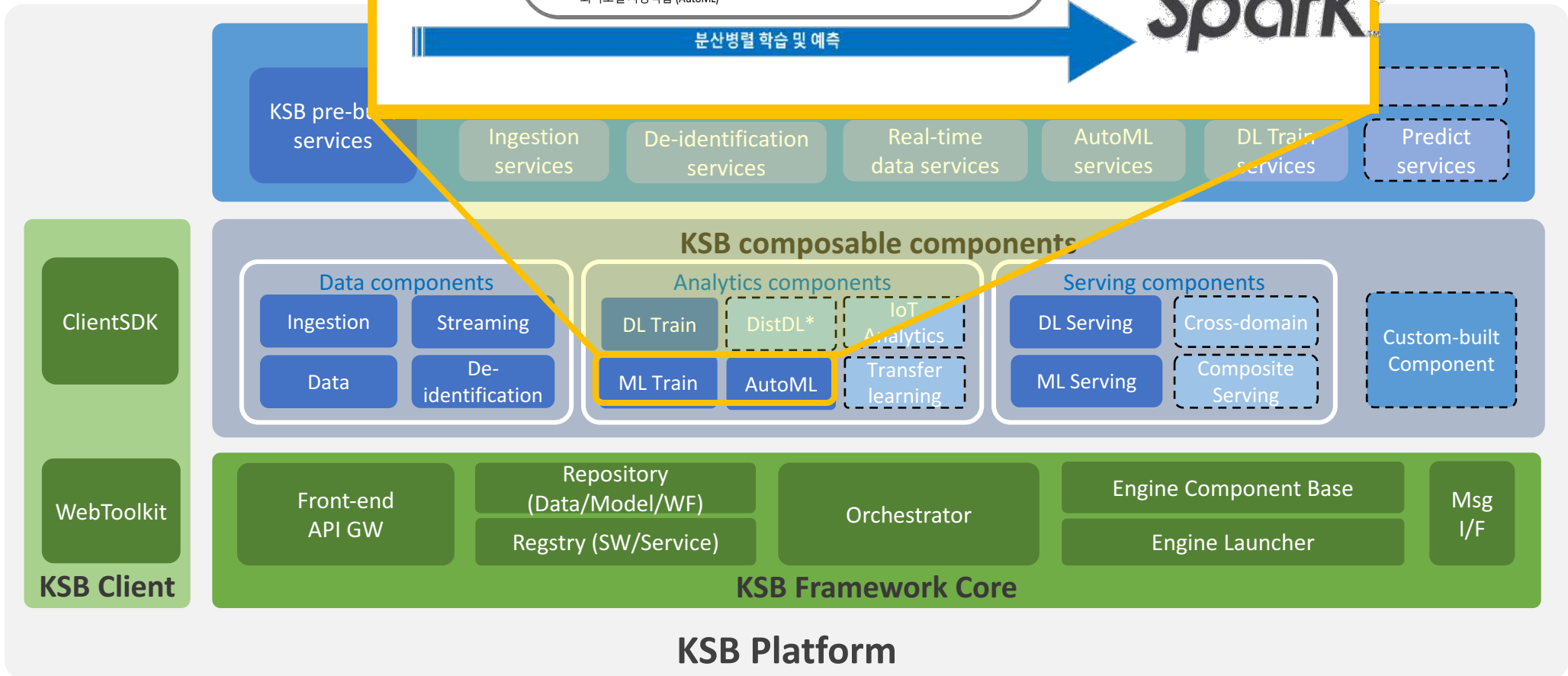
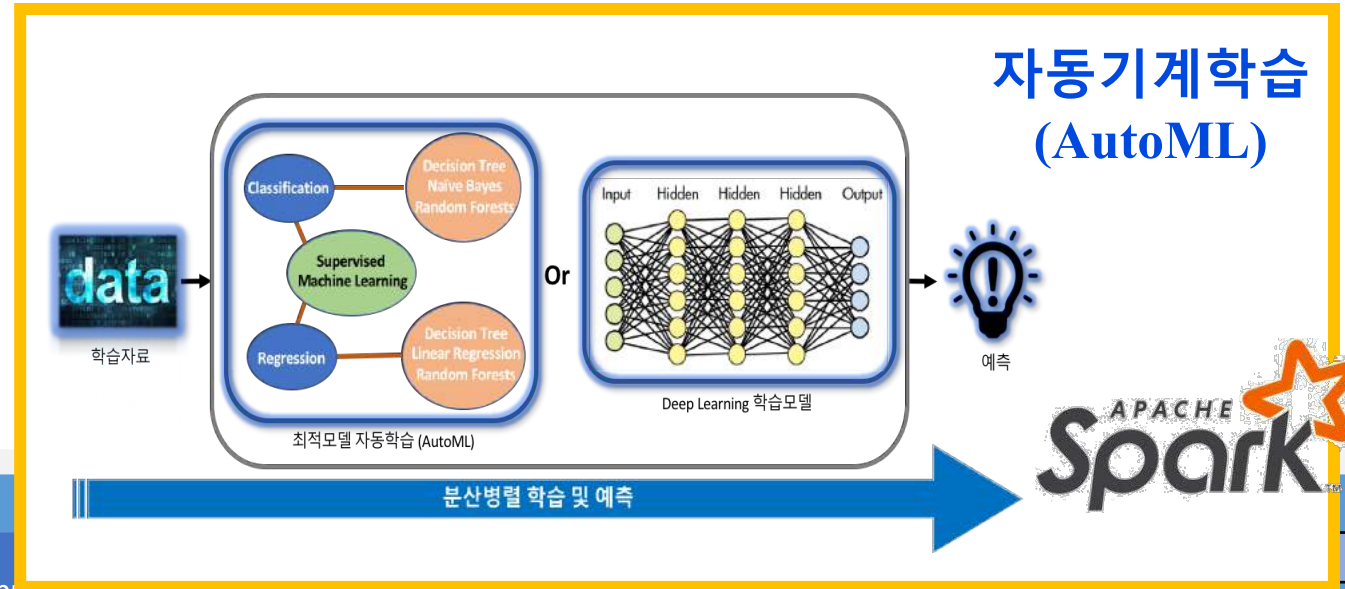
WebToolkit

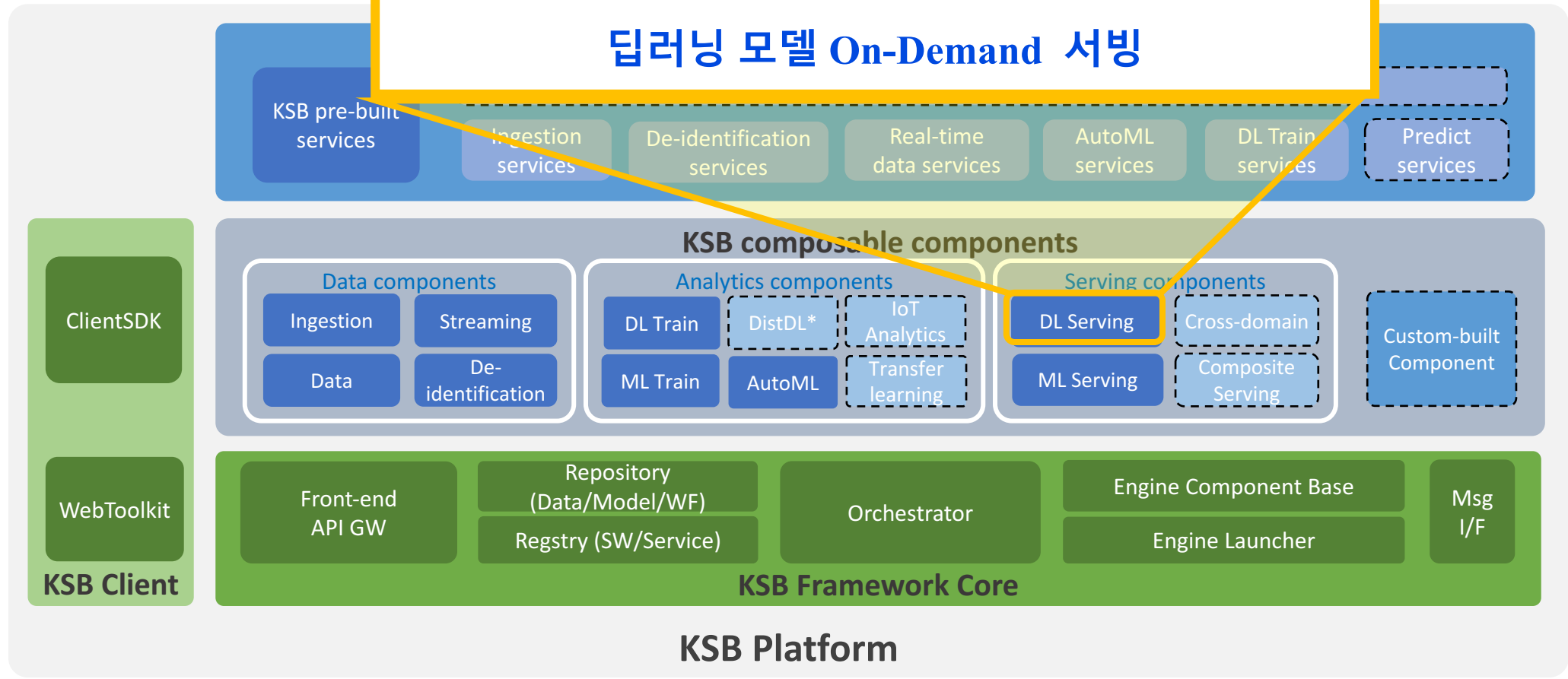
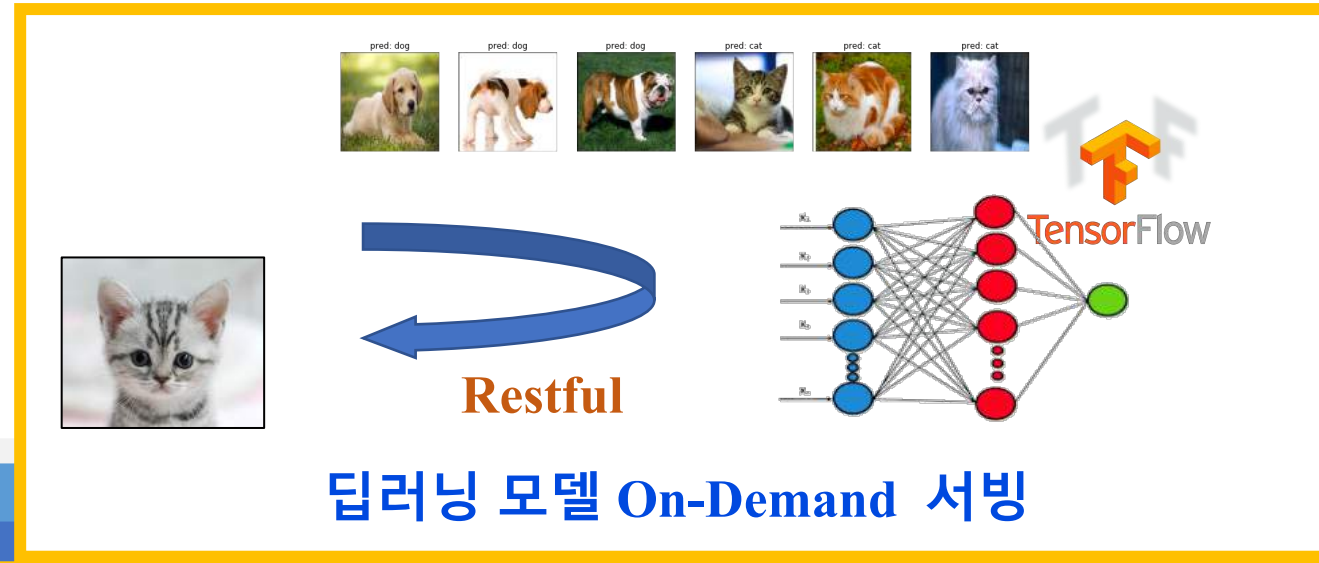
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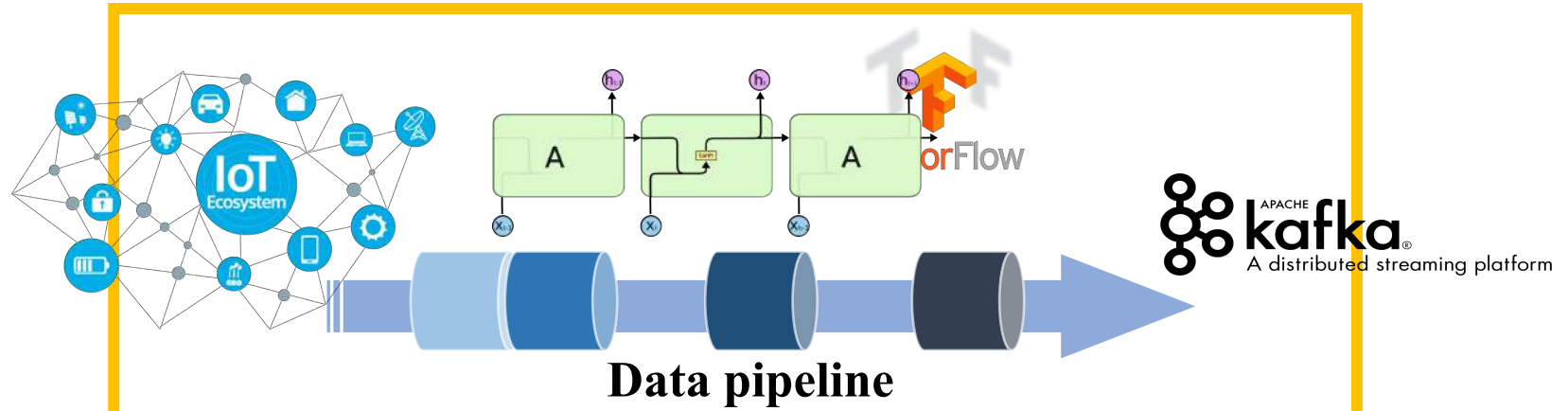


KSB Framework Core

KSB Platform







딥러닝 모델 스트리밍 서빙

KSB pre-built services

Ingestion services

De-identification services

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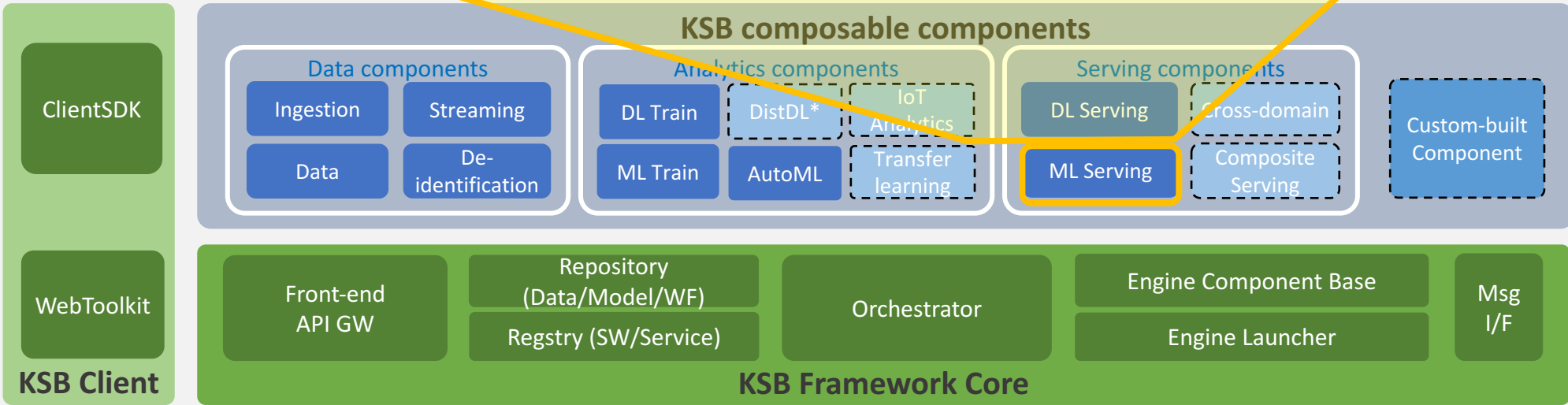
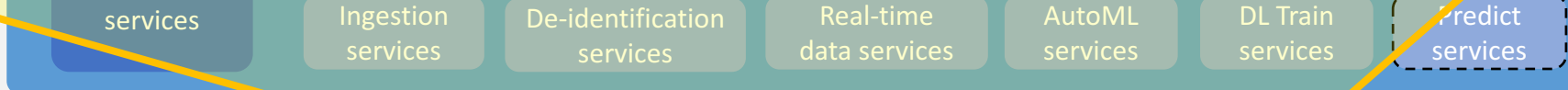
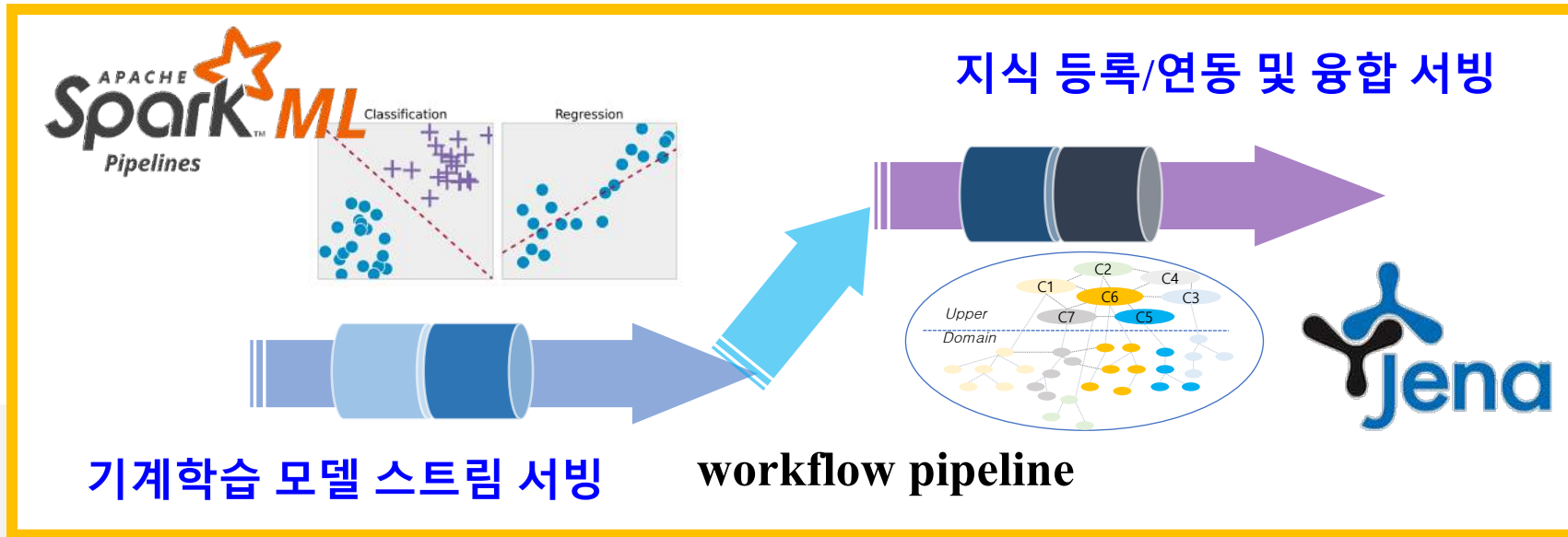
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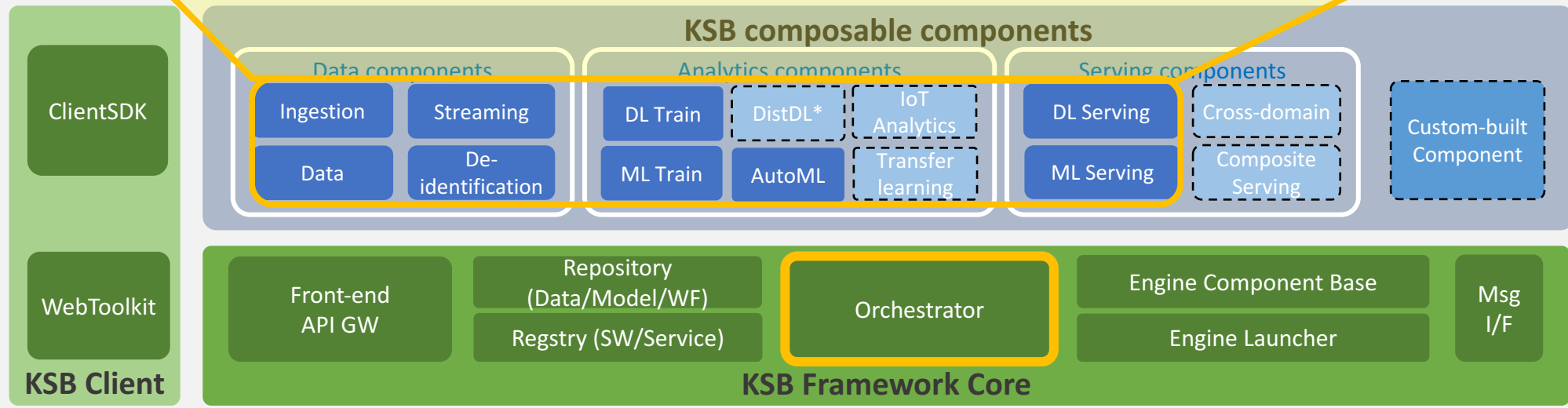
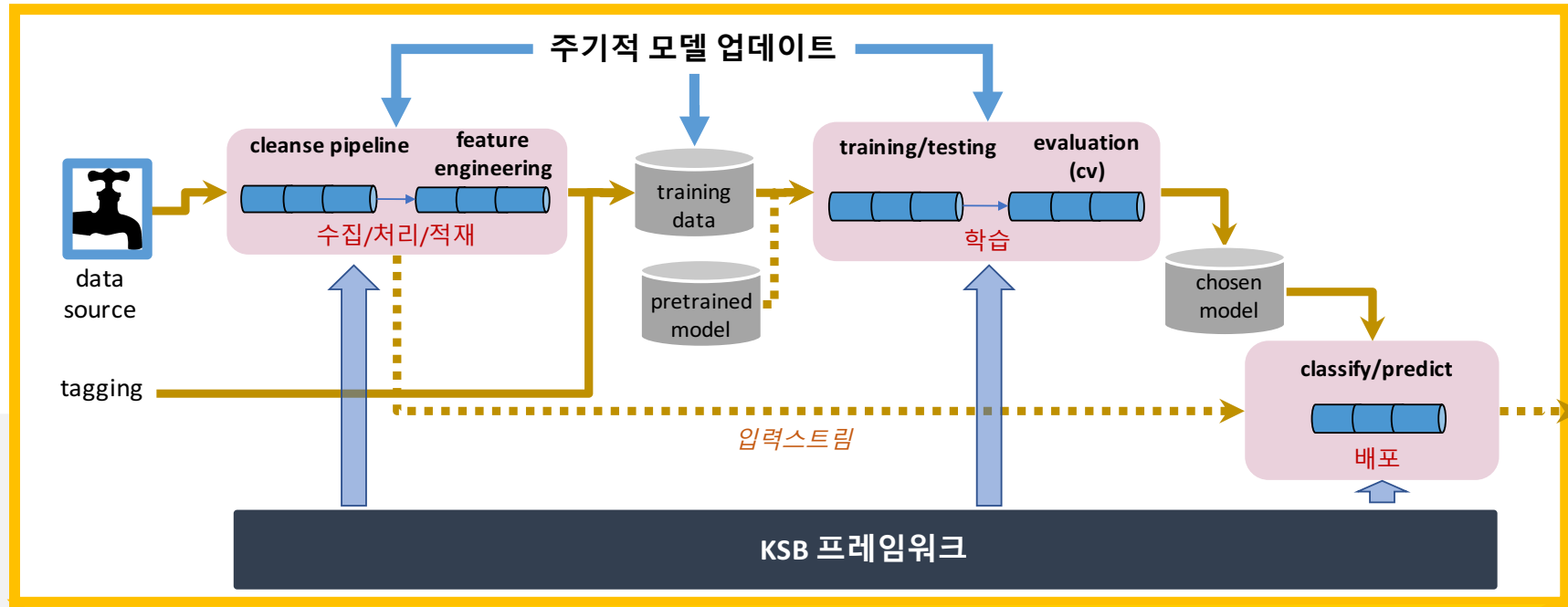
Msg I/F

KSB Framework Core

KSB Platform



KSB Platform

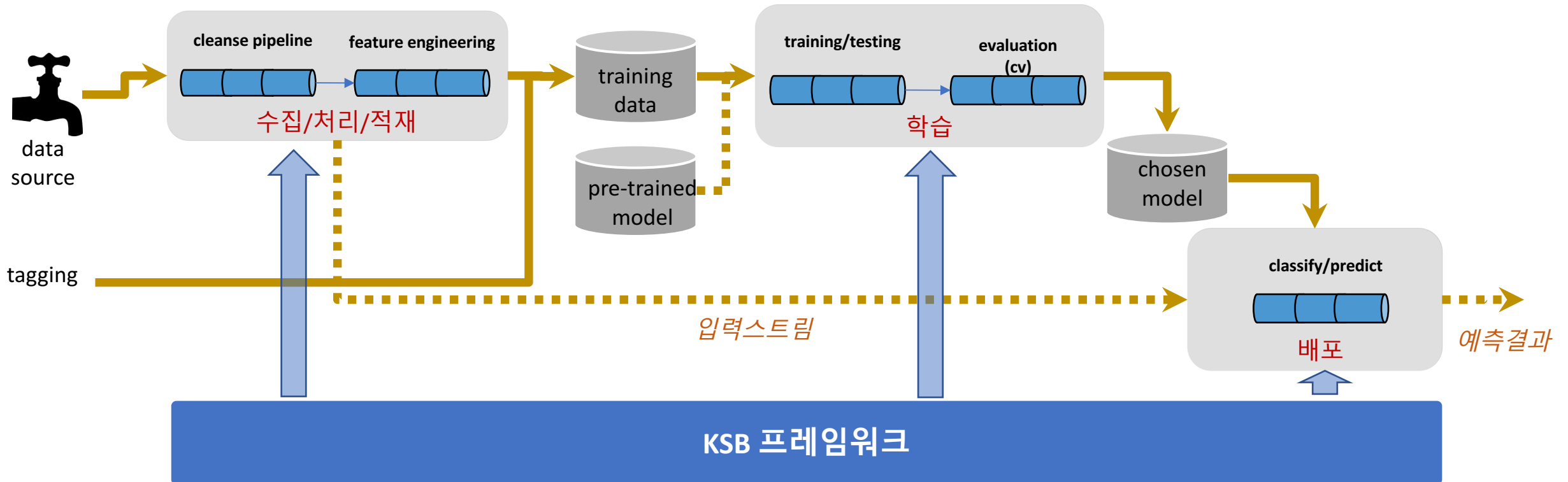


KSB Platform

KSB Framework Key Feature

- End-to-end IoT Intelligence
 - From real-time streaming to streaming predictive serving
 - Predictive serving API to domain application serving API
- Easy of Use
 - DIY Workflow Editor
 - Develop and Deploy in One Place
- Extensibility
 - Accumulating SW component
 - Building Predictive Api using ML Model

End-to-end Machine Learning Workflow

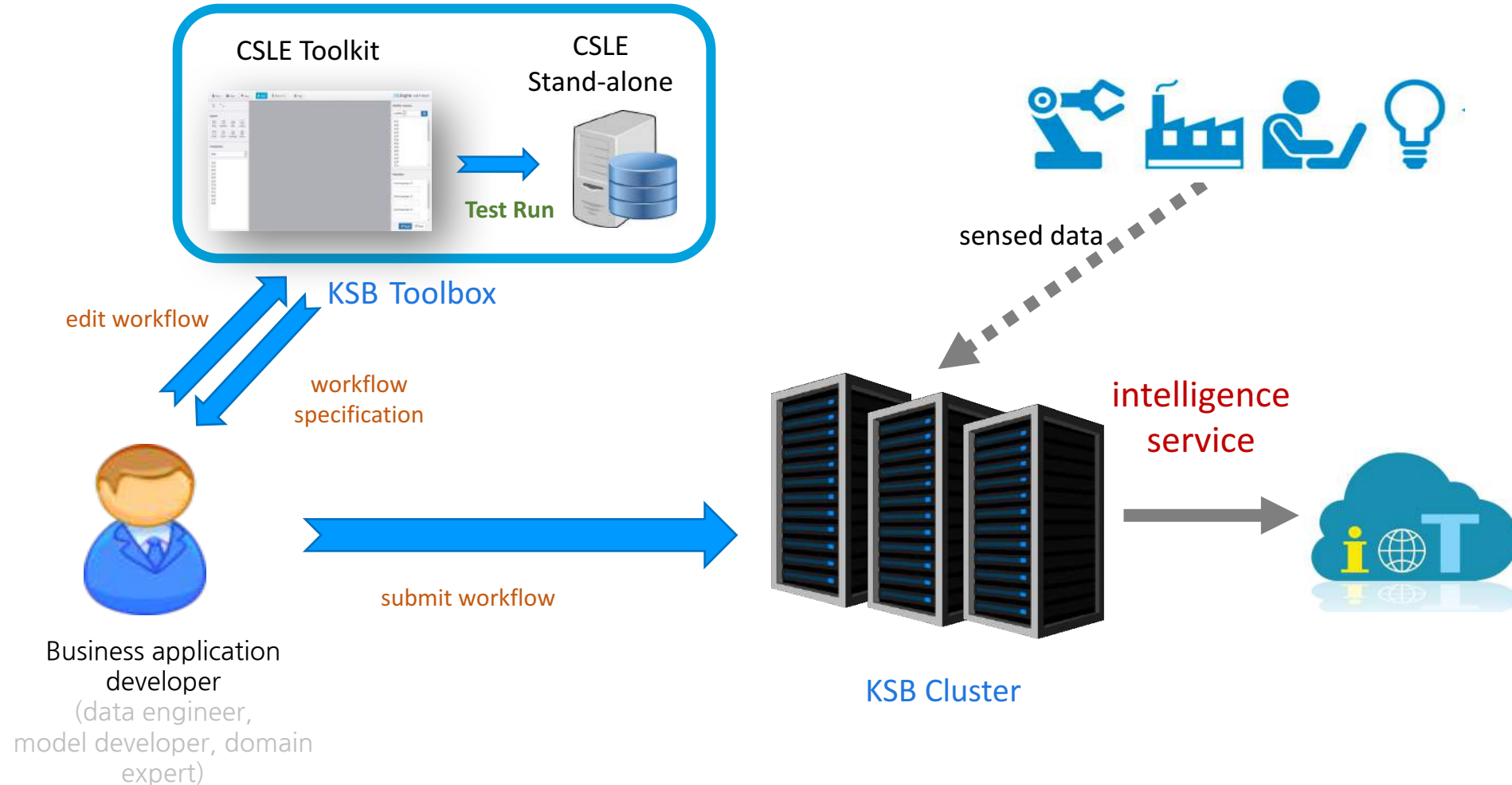


Easy of use : DIY Workflow Editor

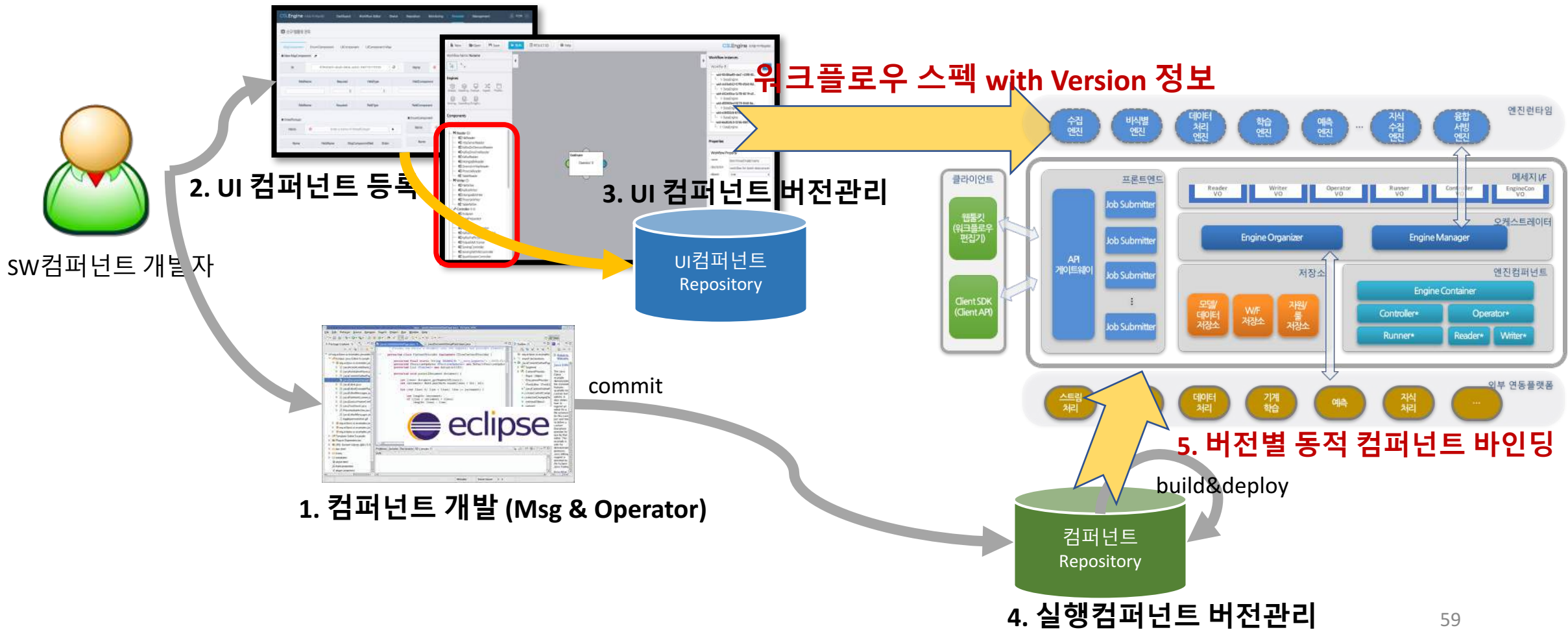
The screenshot displays the Workflow Editor interface, which is used for designing and managing workflows. The interface is divided into several main sections:

- Workflow Name:** Noname
- Components:** A list of available components categorized into:
 - Reader (5):** FileReader, HttpServerReader, MongoDBReader, PhoenixReader, TableReader
 - Writer (5):** FileWriter, MongoDBWriter, PhoenixWrite, StdoutWriter, TableWriter
 - Controller (1):** SparkSessioStreamContr..
 - Runner (3):** PySparkRundener, SimpleSparkRunner, TensorflowRenner
 - Operator (7):** AutoSparkMLOperator, DecisionTreeClassifier, DecisionTreeRegressor, LinearRegressionRegressor, NaiveBayesClassifier, RandomForestClassifier, RandomForestRegressor
- Workflow Canvas:** A central workspace where components are connected. The current workflow consists of:
 - A **BatchDummyEn...** component with a **FileWriter** sub-component.
 - A **BatchDummyEn...** component with sub-components: **Reader**, **PySparkRu...**, **Controller**, **Operator**, and **FileWriter**.
 - A **BatchDummyEn...** component with sub-components: **Reader**, **Runner**, **Controller**, **Operator**, and **FileWriter**.
 - Connections: **FileWriter** (from the first component) connects to **FileReader** (in the second component). **TableReader** (in the second component) connects to **TableWriter** (in the third component).
- Workflow Instances:** A list of running instances, including:
 - aggClustering_20180105 (1:StreamtobatchEnging)
 - aggClustering_201806 (1:DataEngine)
 - aggClustering_201806 (1:StreamtobatchEnging)
 - aggClustering_201806 (1:StreamtobatchEnging)
 - aggClustering_201806 (1:DataEngine)
 - aggClustering_201806 (1:DataEngine)
- Properties:** A panel for configuring workflow properties, including:
 - Workflow Properties:** Name, Description, isBatch True
- Operators:** A sequence of operators: Auto Spar.. → DecisionTr.. → LinearReg...

Easy of Use: Develop and Deploy in One Place



Extensibility: Accumulating SW component



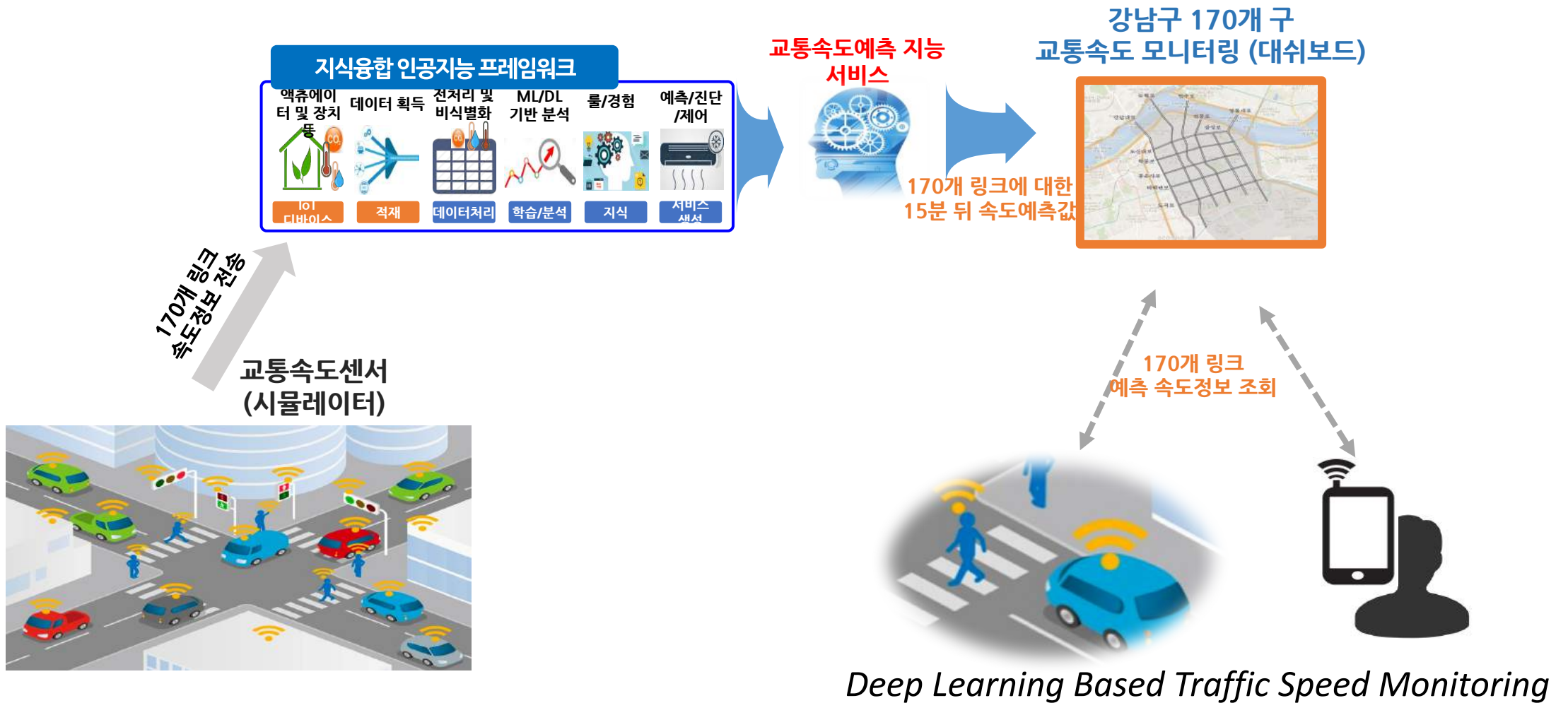
Extendibility: Accumulating SW component

The screenshot displays a 'Workflow Editor' interface with the following components and annotations:

- Workflow Name:** Noname
- Components Panel (Left):** A tree view of components categorized into Reader (5), Writer (5), Controller (1), Runner (3), and Operator (7). A red box highlights this list, with the Korean word '축적' (Accumulation) written below it. An orange arrow labeled '활용' (Usage) points from this list to a component in the workflow.
- Workflow Canvas (Center):** A grid-based workspace containing several 'BatchDummyEn...' components. Each component has a configuration panel with options like Reader, PySparkRu..., Controller, Operator, and FileWriter. Connections between components are labeled 'FileWriter', 'TableReader', and 'TableWriter'. An orange arrow labeled '활용' (Usage) points from a component in the canvas to the 'Workflow Instances' panel.
- Workflow Instances Panel (Right):** A list of workflow instances, including 'aggClustering_20180105' and 'aggClustering_201806'. A red box highlights this panel, with the Korean word '축적' (Accumulation) written next to it. Below it is a 'Properties' section with fields for Name, Description, and isBatch True.
- Operators Panel (Bottom):** A sequence of operators: Auto Spar... → Decision Tr... → LinearReg...

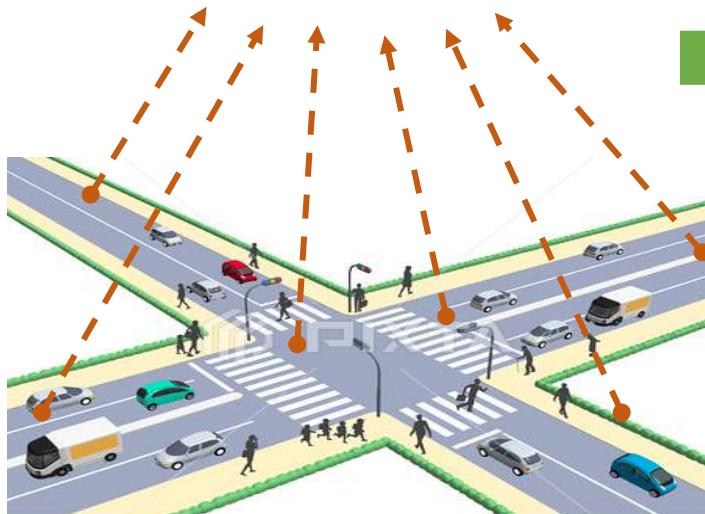
IV. Usecases

U1. Real-time streaming analytics with continuous model update



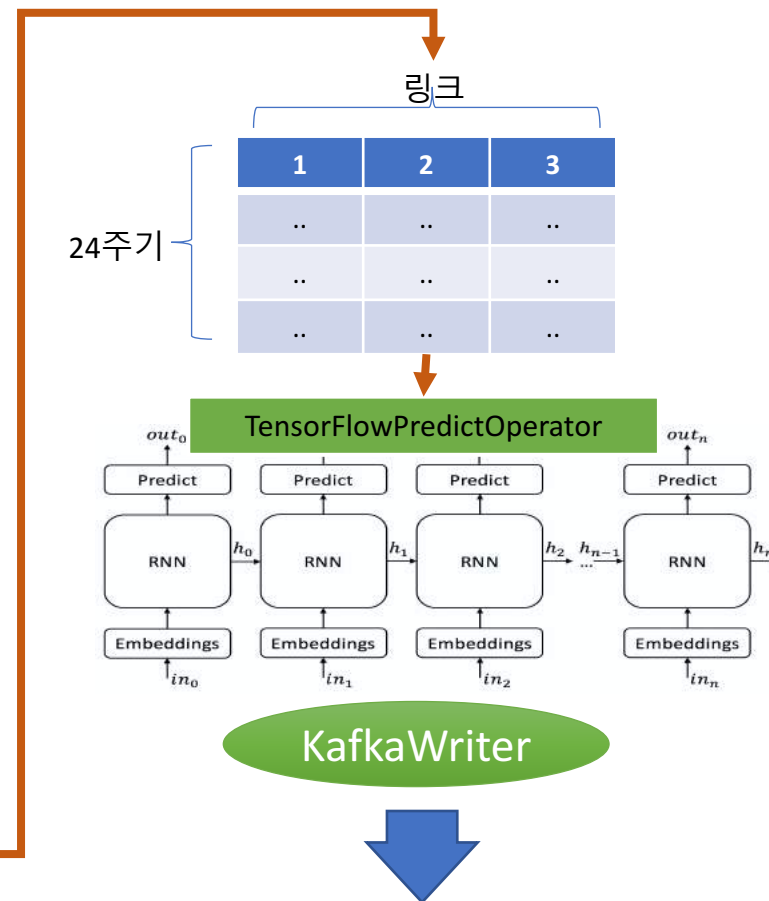
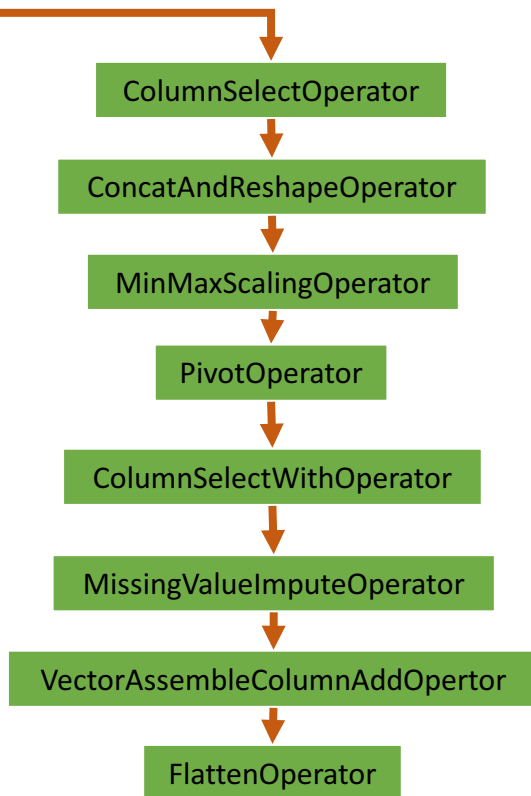
시간	ID	속도
..
..
..

httpReader



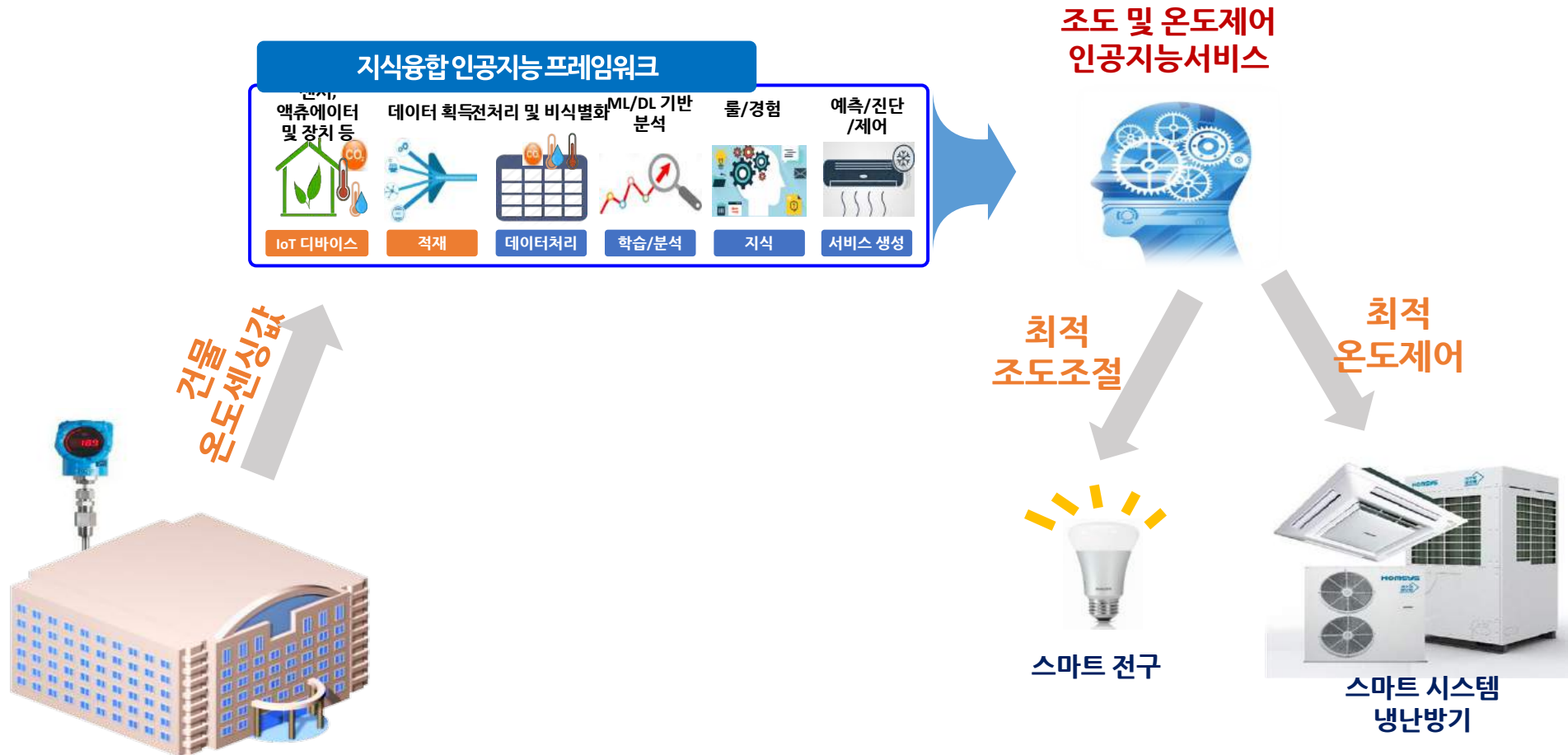
pixtastock.com - 21291829

1,440개 교통정보 수집 센서

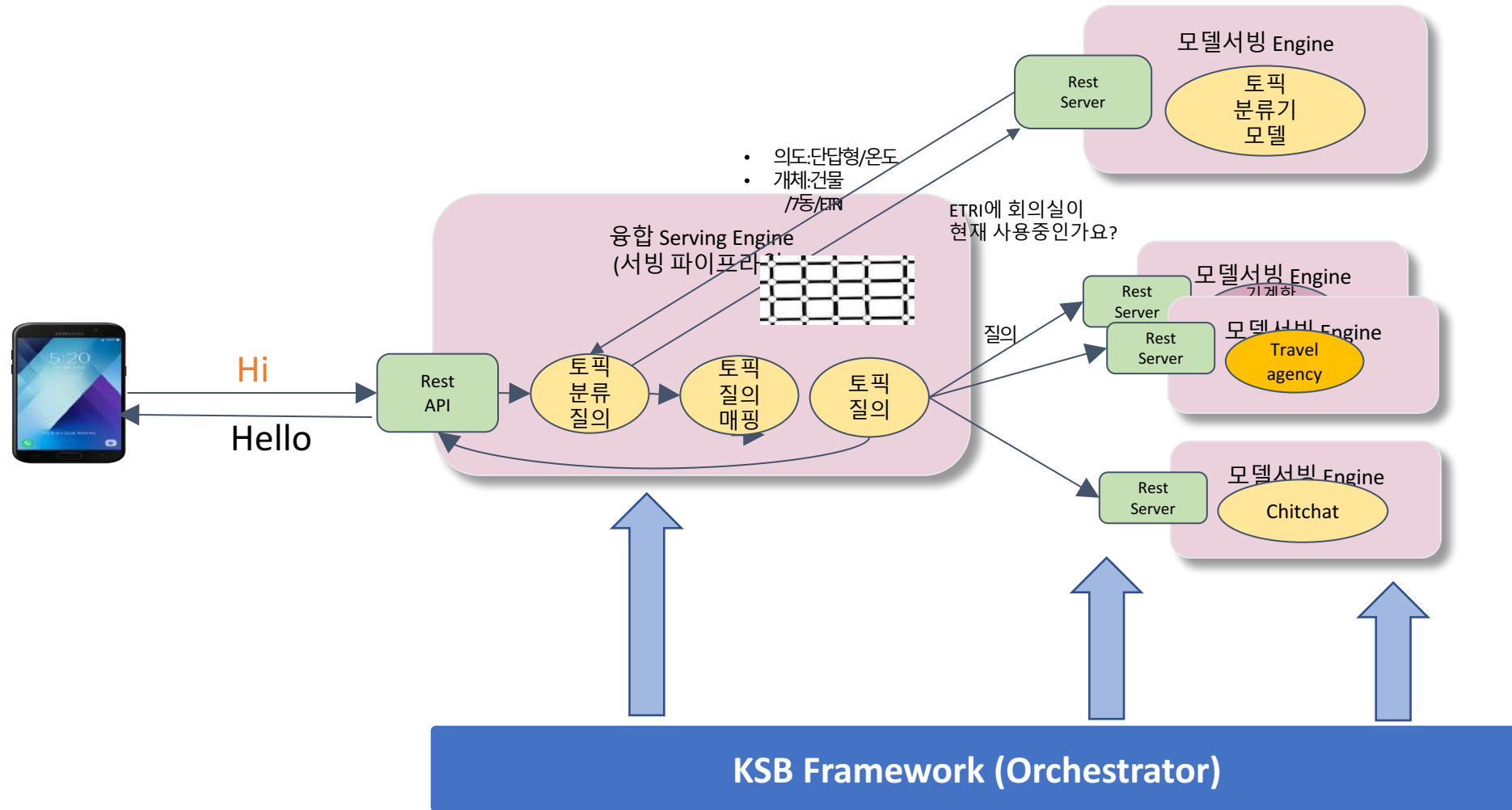


예측/모니터링

U2. Energy efficient building control (Prototype)

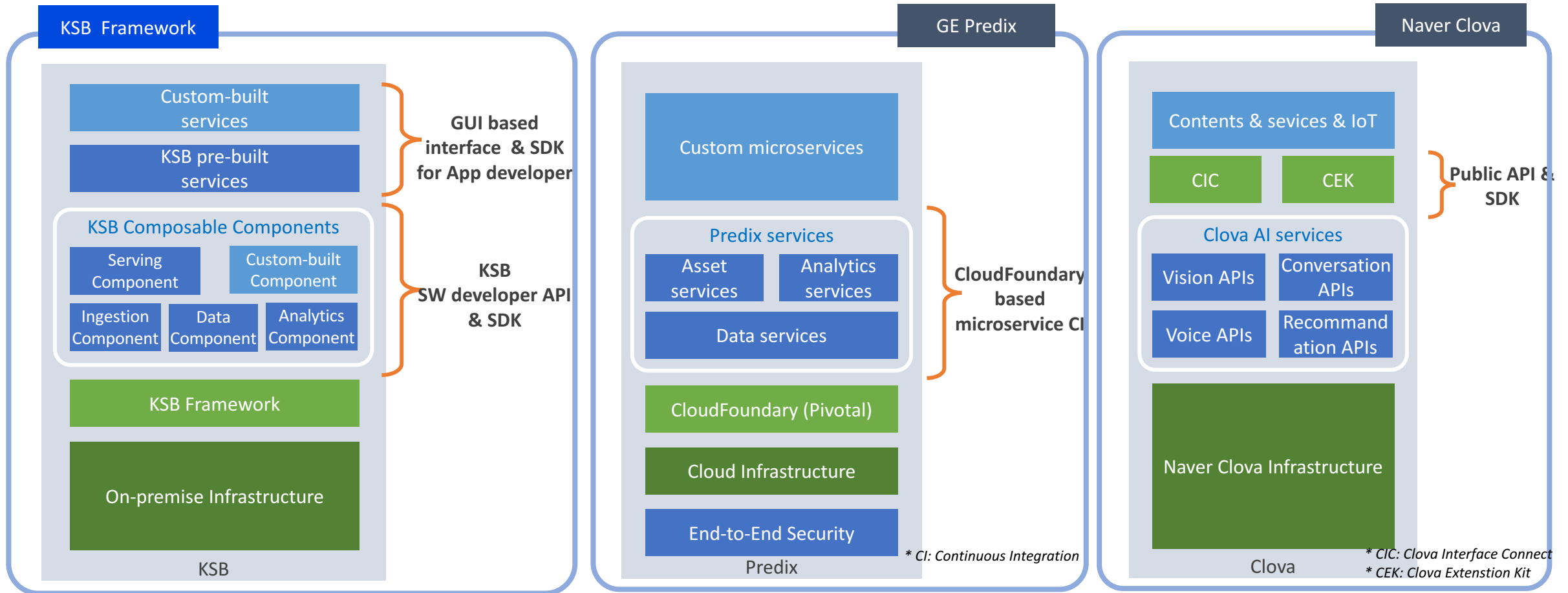


U3. Chatbot (prototype)



V. Conclusion

KSB vs. AI Platforms (Predix/Clova)



Conclusion

KSB VISION is to create Intelligence information industry ecosystem and to spread technology



공공분야 융합 선도 서비스 구축

Conclusion

기업의 경험 지속적으로 축적하고,

공유 및 협력을 위한 장과,

기술의 장벽을 넘어 끊임없이 진화할 수 있는 기반기술 제공

Thank you !