



Practical and Regulatory Considerations for Machine Learning Models Applied to Process Development and Control

Ben Stevens

Disclaimer

Ben Stevens is a current employee of the GSK group of companies and holds shares in GSK.

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Agenda



- 1. Background and Regulatory Context**
- 2. GSK Case Study and Discussions with QIG**
- 3. Other Regulatory Considerations for AI/ML**

Background - Model Classification

Data-Driven/Empirical

Feature:

- Based on **data-driven observations** and used to model the relationship between the system input and output variables.
- Can be useful for **complex systems** and typically **requires minimal understanding** of the science governing the system.
- These models should not be extrapolated beyond the ranges covered by the input data.

Example:

- Multivariate models
- Regression models
- Neural networks

Hybrid (Semi-Empirical)

Feature:

- **Combine empirical and mechanistic** to describe a well-understood part of a system to build a mechanistic model, and where there is a gap or less clearly understood aspect of a system, empirical models can be developed.
- Predictive within the experimental ranges where its empirical part was calibrated but has the advantage of still providing a physical interpretation due to its mechanistic part.

Example:

- Scale-up models using fundamental relations of a system, combined with data-driven experimental data.

Knowledge-Driven/Mechanistic

Feature:

- Based on understanding the science governing the system and used to **model the underlying phenomenon** of a system and its relationship to the output.
- Can perform **predictions beyond the ranges covered (extrapolation) by the input data** (depending on the validity of the underlying assumptions).

Example:

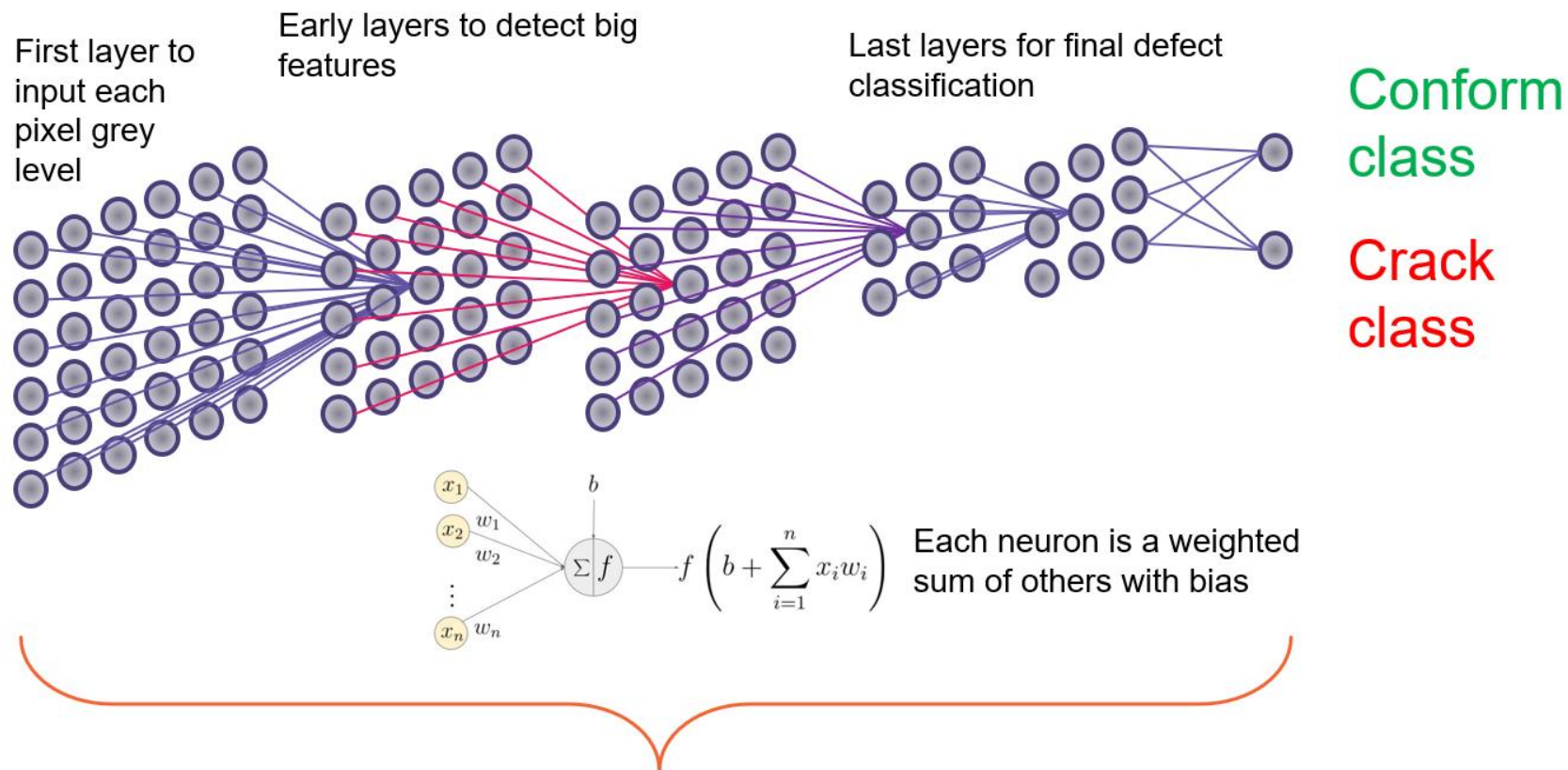
- Chemical Kinetics Models
- Population balance model (PBM)
- Computational Fluid Dynamics (CFD)

Deep Learning for Automated Visual Inspection

Conform images

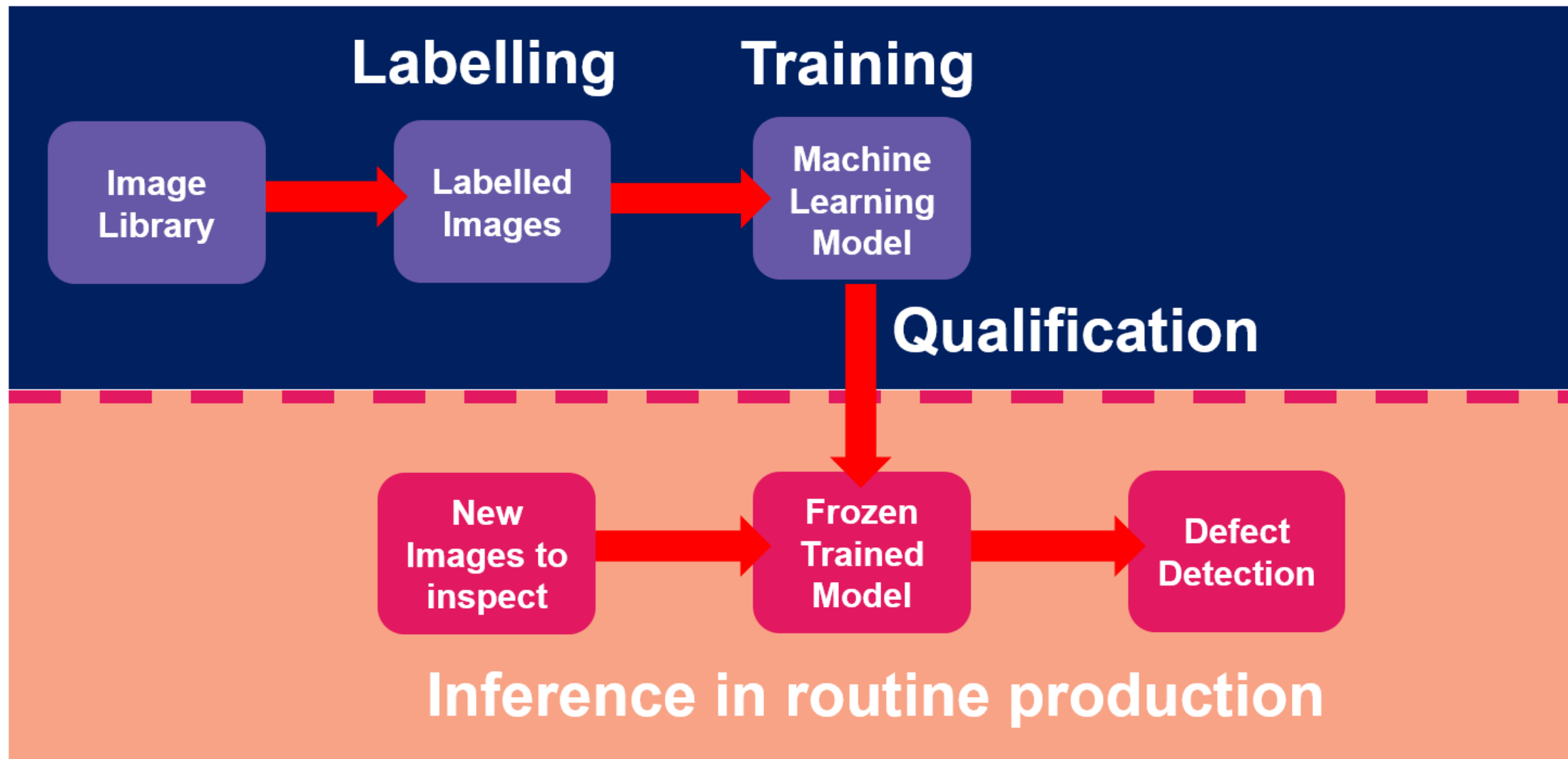


Crack images

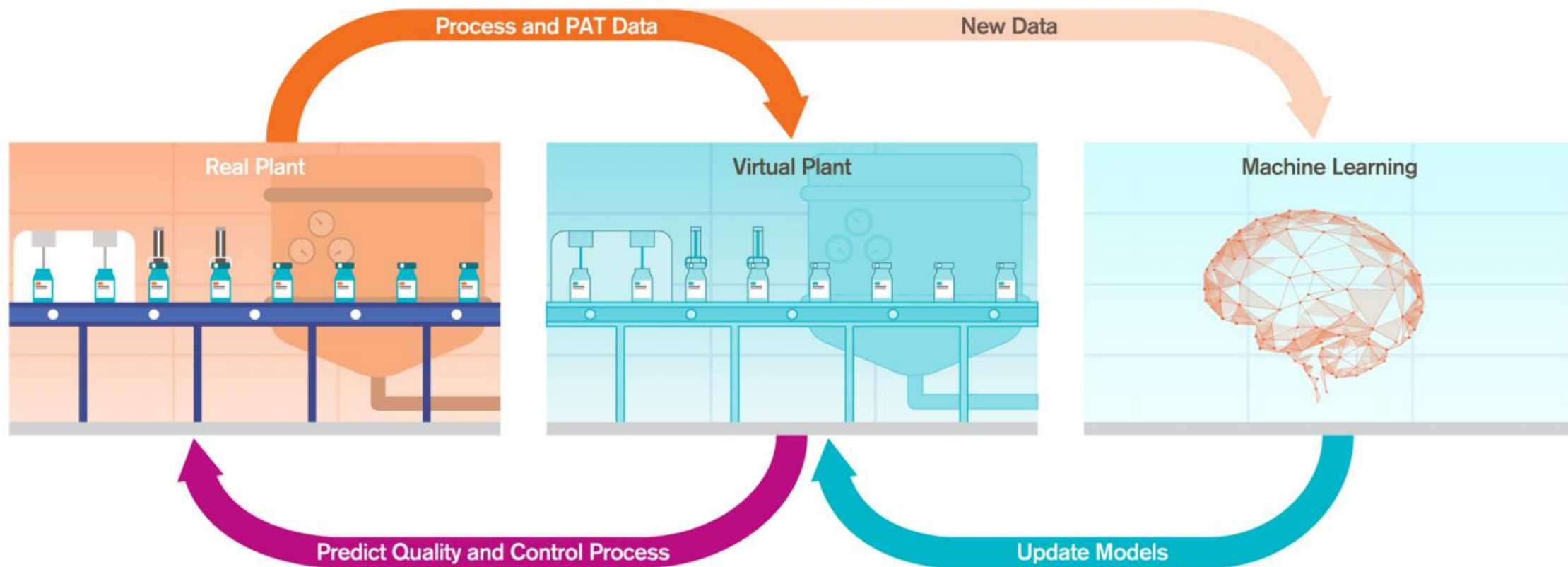


Many Layers designed to optimize image classification, containing from 3 to 50 million parameters to adjust

Deep Learning for Automated Visual Inspection



Hybrid Process Models – Digital Twins



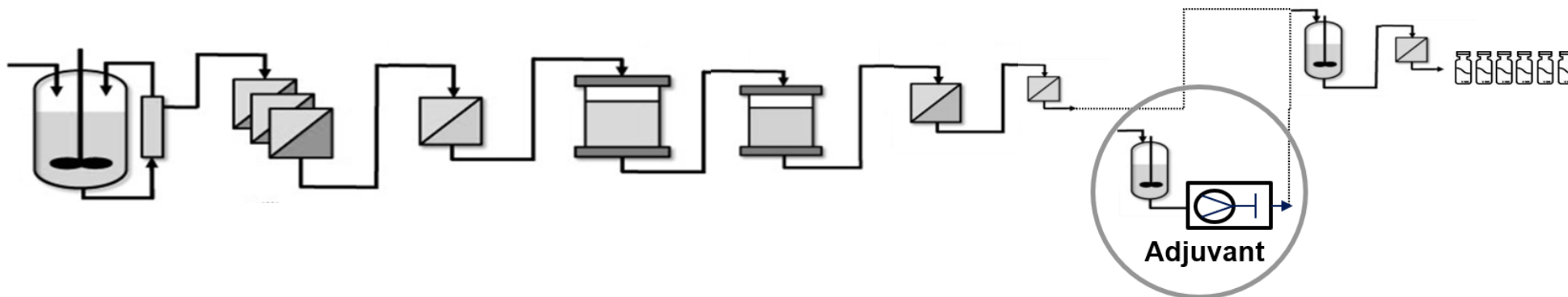
Online: Assurance of quality

Collect process data in real time, understand what is happening and provide optimal control

Offline: Accelerated development

Do in-silico development: simulate, test, optimize before experimenting in the lab

Digital Twin for Vaccine Adjuvant Manufacture



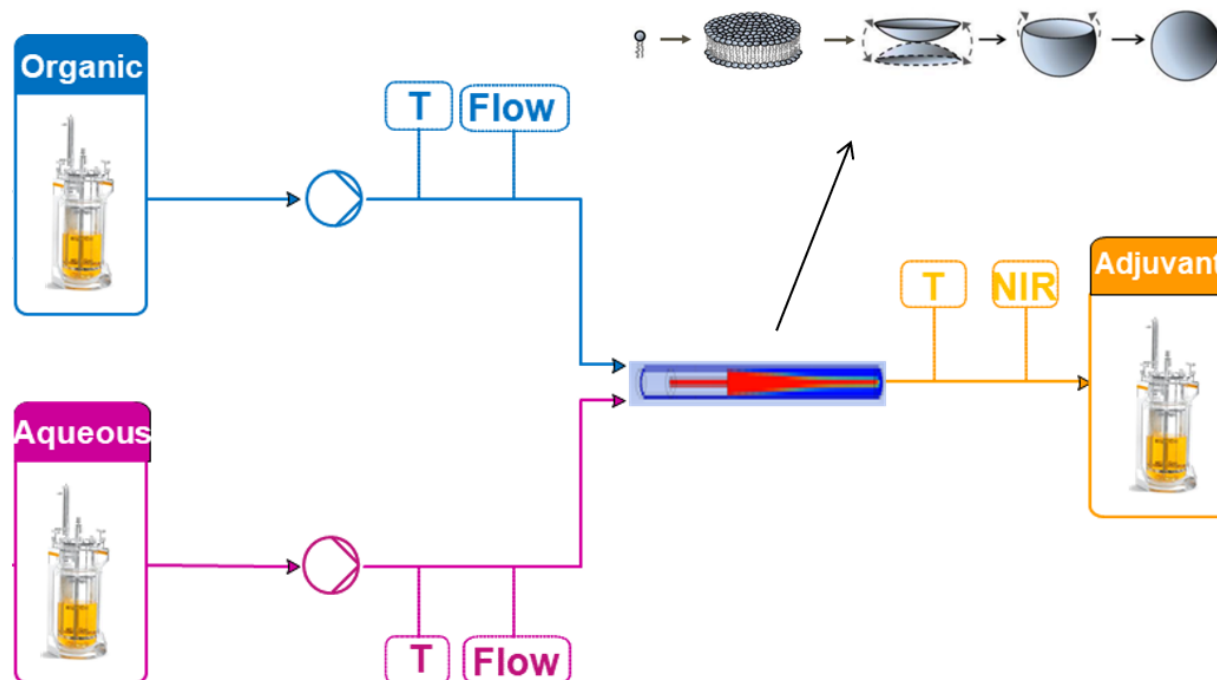
Critical Process Parameters

- Flow rates
- Concentrations
- Temperature

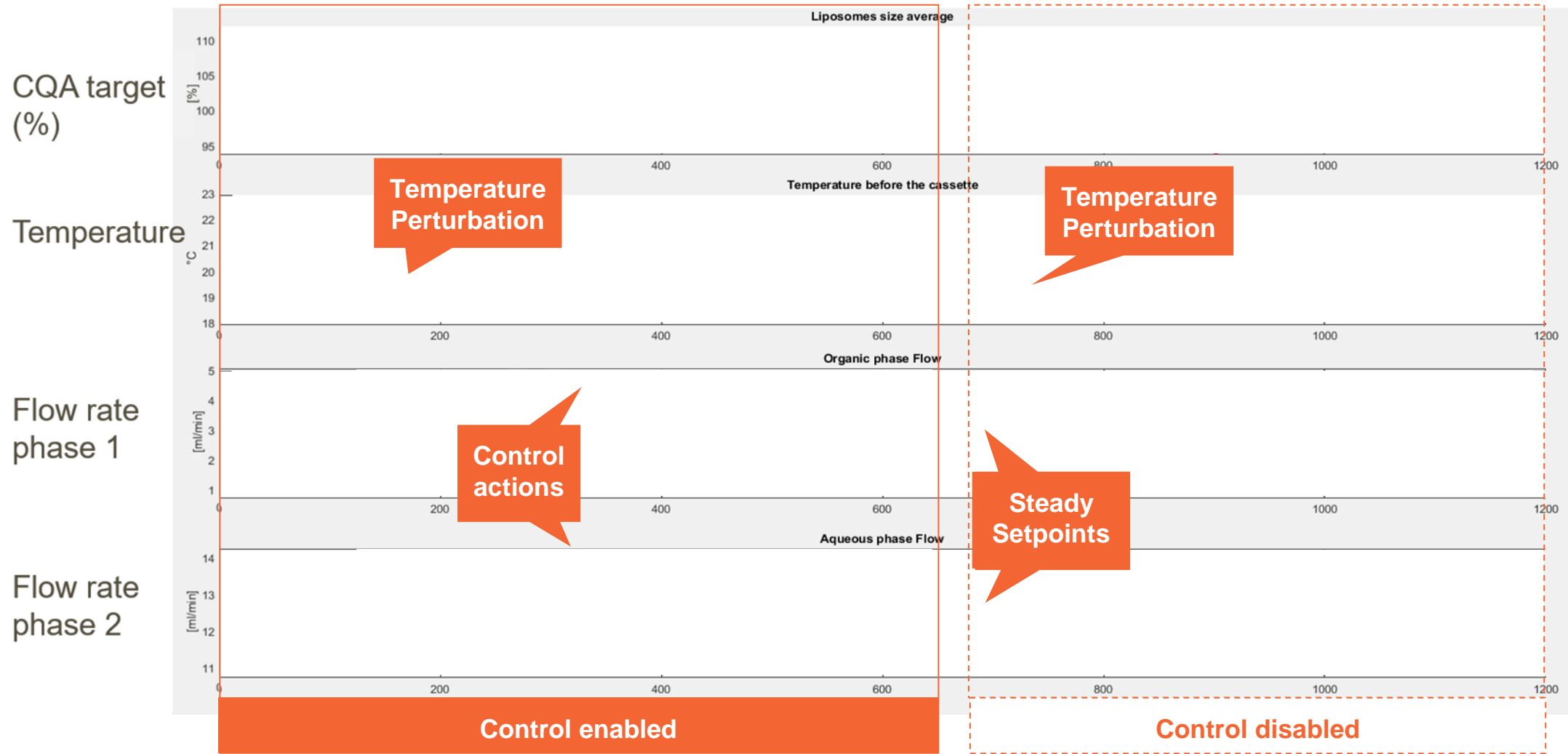
MODELS to PREDICT & ACT

Quality Attributes

- Adjuvant concentration
- Adjuvant size distribution



Digital Twin for Vaccine Adjuvant Manufacture



Regulation is Here... but Rapidly Evolving

OCTOBER 30, 2023

FACT SHEET: President Biden Issues Executive Order on Safe, Secure, and Trustworthy Artificial Intelligence

 BRIEFING ROOM ▶ STATEMENTS AND RELEASES

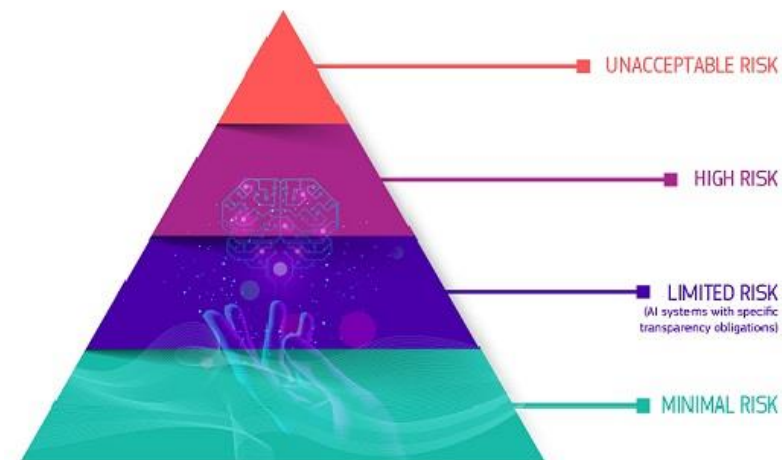
- “The Executive Order establishes new standards for AI safety and security, protects Americans’ privacy, advances equity and civil rights, stands up for consumers and workers, promotes innovation and competition, advances American leadership around the world, and more.”

<https://www.whitehouse.gov/briefing-room/statements-releases/2023/10/30/fact-sheet-president-biden-issues-executive-order-on-safe-secure-and-trustworthy-artificial-intelligence/>

- “At least 12 (states) have enacted laws that delegate research obligations to government or government-organized entities to increase institutional knowledge of AI and better understand its possible consequences.”

<https://www.brennancenter.org/our-work/research-reports/states-take-lead-regulating-artificial-intelligence>

• EU AI Act:



STEP 1



A high-risk AI system is developed.

STEP 2



It needs to undergo the conformity assessment and comply with AI requirements.*

*For some systems a notified body is involved too.

STEP 3



Registration of stand-alone AI systems in an EU database.

STEP 4



A declaration of conformity needs to be signed and the AI system should bear the CE marking. **The system can be placed on the market.**

If substantial changes happen in the AI system's lifecycle

GO BACK TO STEP 2

<https://www.consilium.europa.eu/en/press/press-releases/2023/12/09/artificial-intelligence-act-council-and-parliament-strike-a-deal-on-the-first-worldwide-rules-for-ai/>

Regulation is Here

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REVOKED

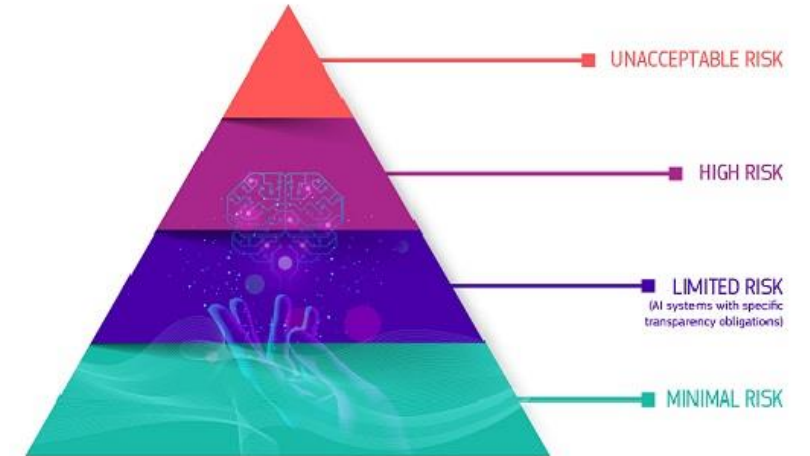
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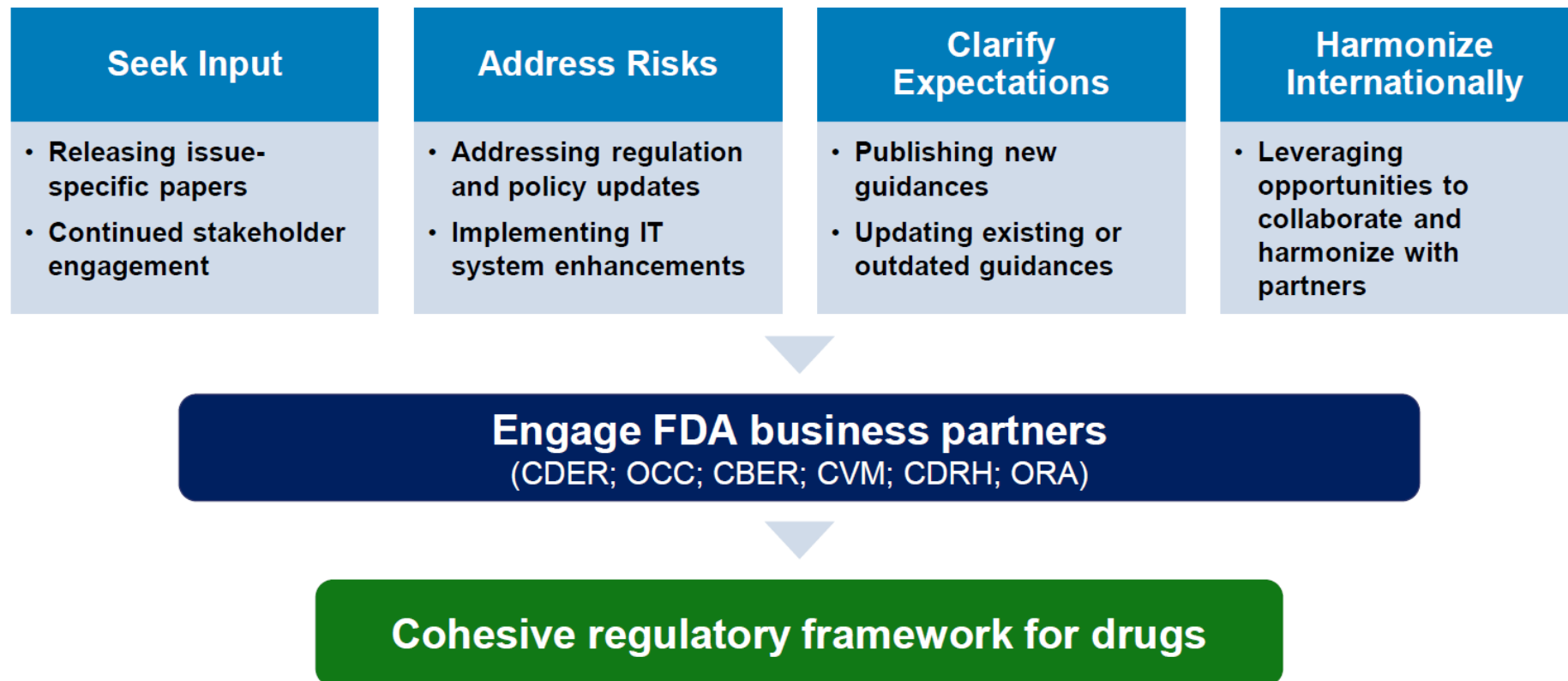
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Key Developments: FDA Framework for Regulatory Advanced Manufacturing Evaluation (FRAME)

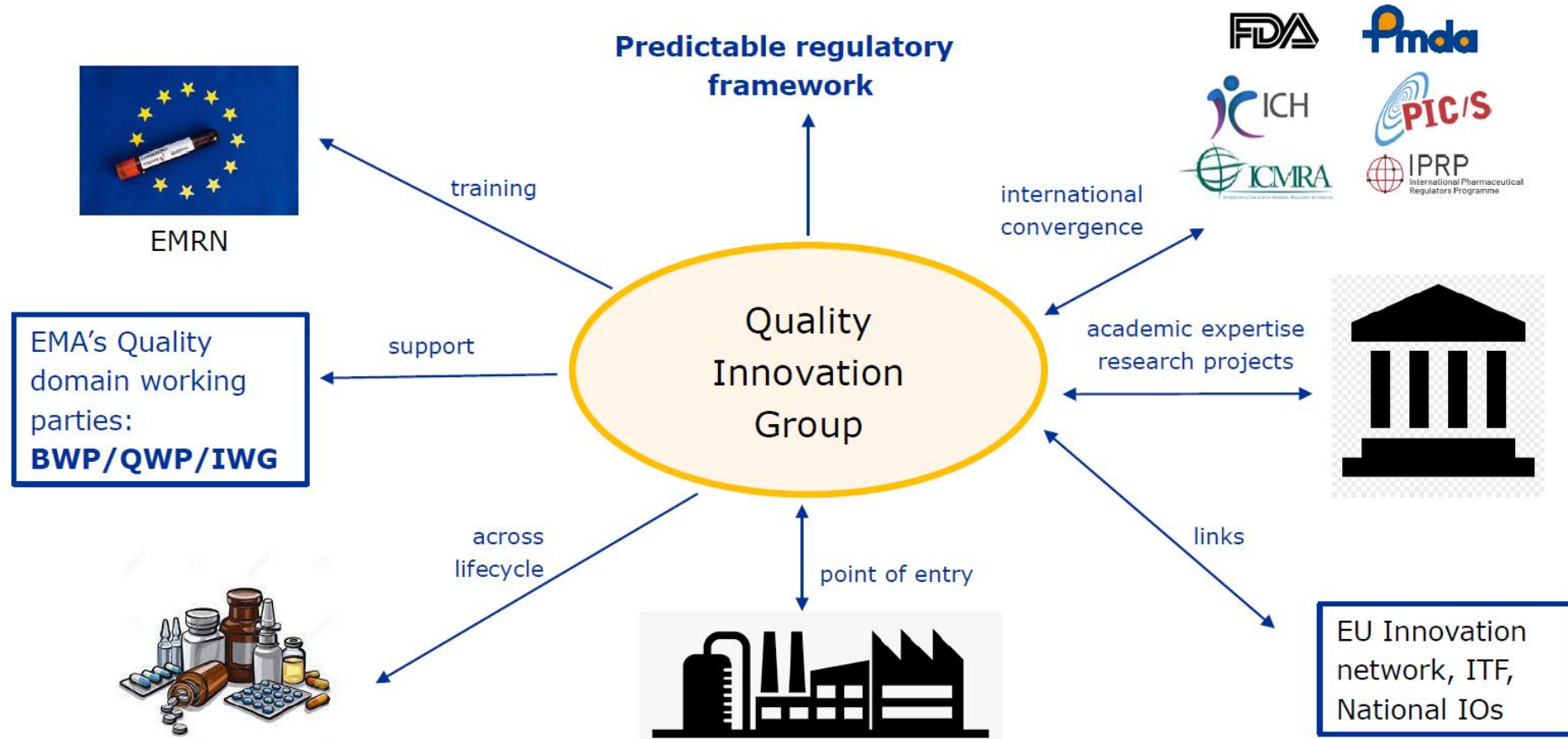


FDA's FRAME provided important 2023 concept paper for AI/ML in drug manufacturing and sponsored a critical dialogue through PQRI Workshop.

<https://www.fda.gov/about-fda/center-drug-evaluation-and-research-cder/cders-framework-regulatory-advanced-manufacturing-evaluation-frame-initiative>

Key Developments: EMA Quality Innovation Group (QIG)

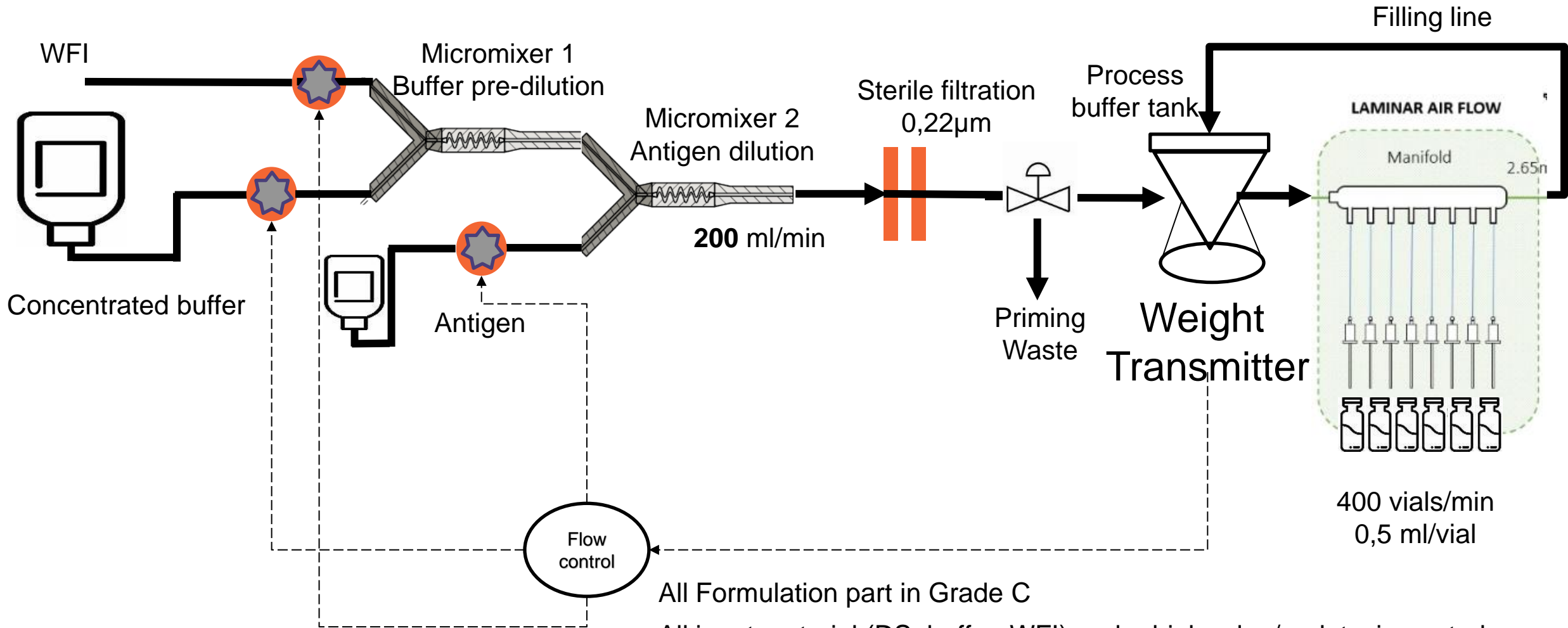
2023 EMA QIG Digital Listen and Learn provided a critical forum for discussion of AI/ML use in process modeling and GMP applications.



<https://www.ema.europa.eu/en/committees/working-parties-other-groups/chmp/quality-innovation-group>

https://www.ema.europa.eu/en/documents/report/report-listen-and-learn-focus-group-meeting-quality-innovation-group_en.pdf

GSK Case Study – Digital Twin for Continuous Formfilling



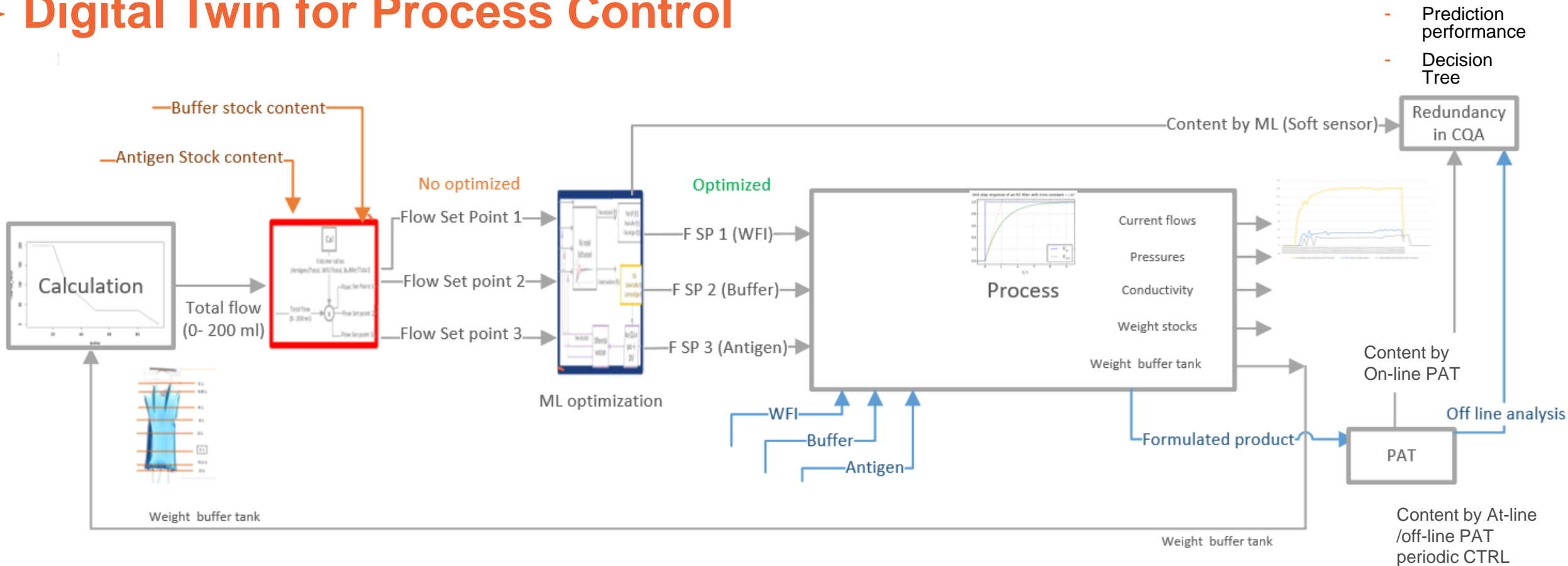
All Formulation part in Grade C

All input material (DS, buffer, WFI) under bioburden/endotoxin control

Single use assembly (bioburden control) including (POFF/PUPSIT)

Close system

Digital Twin for Process Control



- PAT Sensors (conductivity, flow, weight and pressure) and PAT probes (UV and NIR) provide data enabling real-time process monitoring based on Chemometrics models coupled with machine learning models (ML).
- Hybrid system model (“Digital Twin”) capable of simulating time profiles of product content and prediction of other attributes (conductivity, pH, concentration, etc.) from system inputs.
- Direct feedback loop to adjust process parameters to optimize product quality and minimize waste.
- **Full release testing is still carried out (i.e., NOT RTRT)**

ICH Q8/Q9/Q10 Q&A Points to Consider - Model Impact

▪ High-Impact Models:

- Prediction from the model is a **significant indicator of quality** of the product
- Must have high precision and accuracy
- Should be fully **validated** at commercial scale
- Must be **maintained and updated** during the product lifecycle

▪ Medium-Impact Models:

- Useful in **assuring quality** of the product
- **Not the sole indicator** of product quality
- Must have **appropriate precision, accuracy, and predictive power** to assess the probability of failure

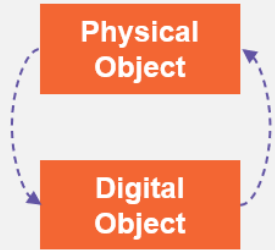
▪ Low-Impact Models:

- **Support** product and/or process development
- Model predictions are **not the direct indicators for assurance of product quality**

GSK Twin Level Definitions

Digital Twin Level 1

(Digital Model)

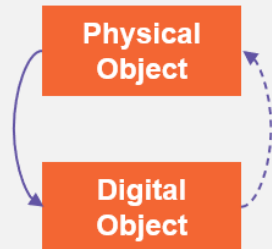


Development

- Reduce experimentation by in silico process development
- Training & process understanding

Digital Twin level 2

(Digital Shadow)

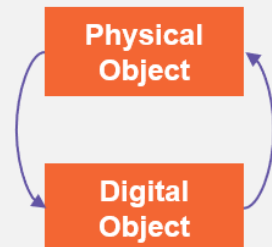


Introduction of new processes

- Provide advanced monitoring
- Recommend action if a trend towards deviation is detected
- CPPs are constrained

Digital Twin level 3

(Digital Twin)



New continuous process & batch processes after learning phase

- Provide advanced monitoring & advanced control to maintain CQAs at target

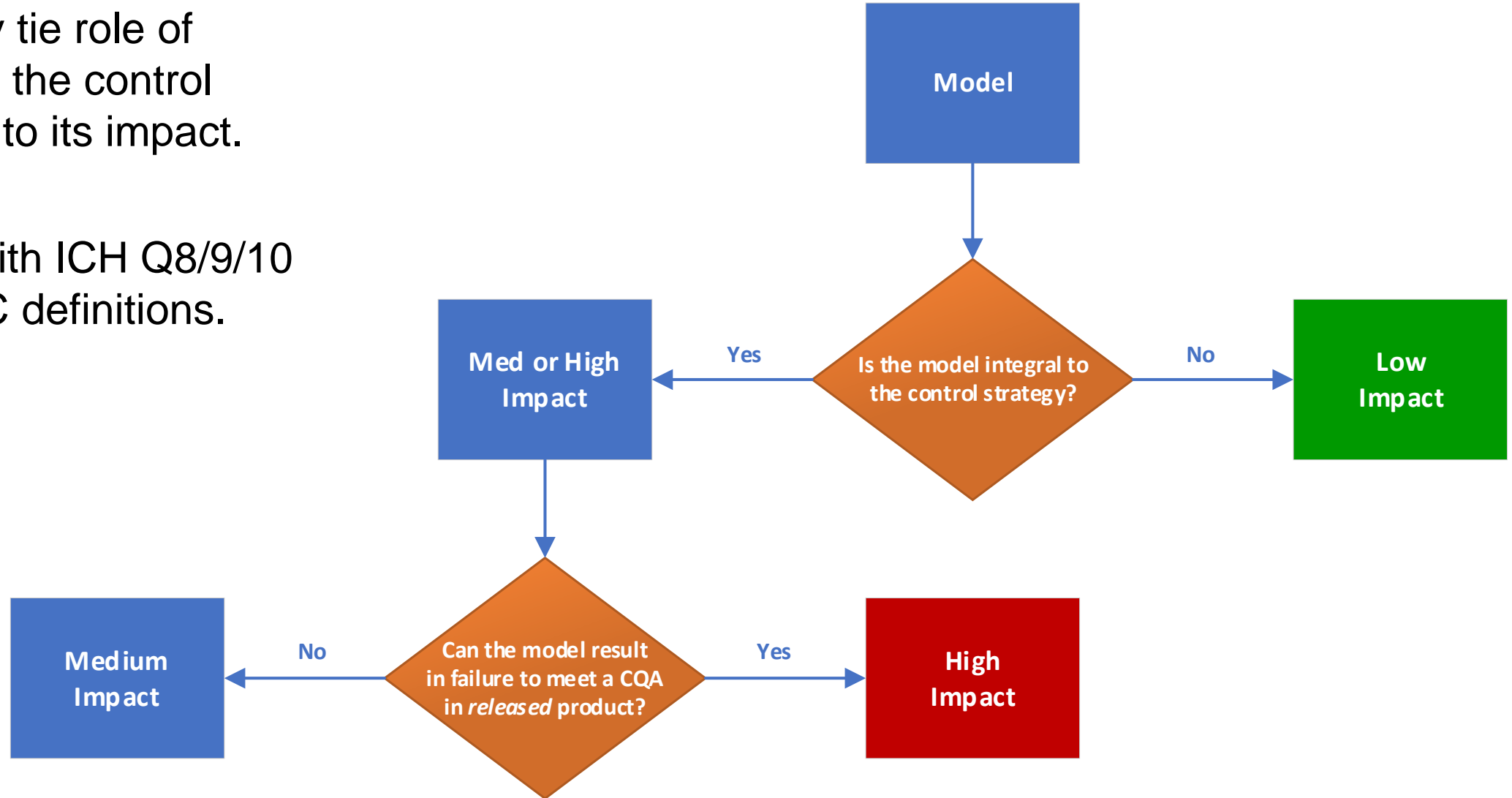
EMA QIG Feedback for GSK

Q: GSK is proposing that for a digital twin model for a continuous process which controls the process but where there is **no decrease in end product testing**, the model will need limited verification at the commercial scale and that model performance can be demonstrated as part of PPQ where superiority of model-based control can be demonstrated over classical (parametric) controls. Is this acceptable to the QIG?

A: The QIG asked GSK to clarify if the proposal is to provide in the application verification elements instead of validation elements. GSK confirmed the understanding of the proposal, indicating small-scale experiments are planned to test the model. For example, by introducing intentional disturbances experiments/simulations to demonstrate that the digital twin could identify, anticipate problems, and adapt accordingly the process. **The QIG agreed that given that the end product testing remains fully in place, the model would be considered low/moderate impact and in level 2, hence this approach should be acceptable.** GSK asked whether this proposal would be acceptable for a level 3-type model as well. **The QIG indicated that if standard QC release is done with no RTRT, this approach can be still acceptable (e.g., the model remains medium impact), provided model performance is appropriately demonstrated by designed small scale or in silico experiments.** The QIG also acknowledged that the digital twin model performance will improve over time as further data is collected. GSK confirmed that model performance will be verified and demonstrated, but not part of formal commercial-scale validation.

General Approach for Process Model Impact?

- Explicitly tie role of model in the control strategy to its impact.
- Aligns with ICH Q8/9/10 Q&A PtC definitions.



ICH Q12 Definitions

Per ICH Q12:

- **Established conditions (ECs)** are legally binding information (or approved matters) considered necessary to assure product quality. As a consequence, any change to ECs necessitates a submission to the regulatory authority.
- A **parameter-based approach** is one in which product development prior to regulatory submission provides a limited understanding of the relationship between inputs and resulting quality attributes and will include a large number of inputs along with outputs.
- A **performance-based approach** is one where ECs are primarily focused on outputs rather than inputs. This is enabled by knowledge gained from an enhanced approach, a data-rich environment, and an enhanced control strategy (e.g., **models**, PAT).

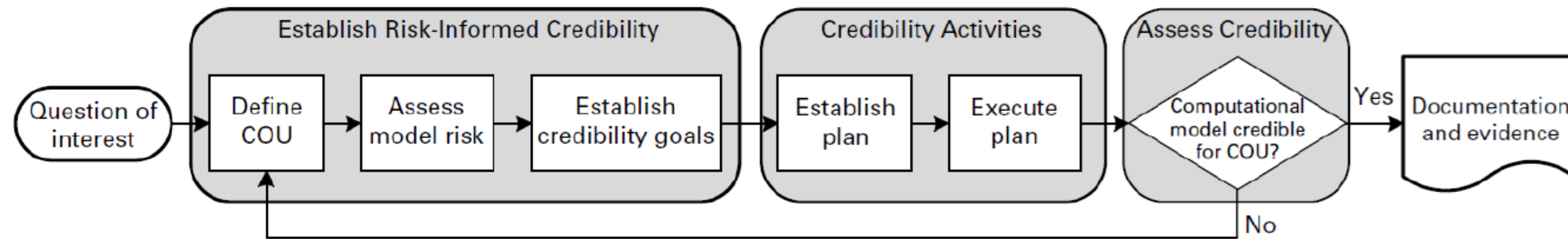
EMA QIG Feedback for GSK

Q: Practical use of a digital twin for process control will mean that the process parameter setpoints adjust automatically, based on the model, within defined ranges. Conceptually, GSK believes this is justifiable based on the overall control strategy, including real-time verification of process outputs, and can be justified in the dossier. However, GSK are concerned that current guidance and requirements regarding “design space” (or moreover EMA expectations for parameter ranges/PARs) do not fully anticipate the envisioned scenario. Narrow interpretation and strict application of these design space guidelines could inhibit implementation and use of these models. **Can the framework described in ICHQ12 Section 3.2.3.1 for a “performance based” process control strategy be applied, such that the manufacturing process is not described by process parameter ranges?**

A: **QIG indicated that performance-based process control strategy per Q12 (i.e., one not described by fixed parameter ranges, but relies on the controls of the model) is recognized.** The QIG indicated that, unlike mechanistic or metabolic models, truly data driven models may not be fully understood. The QIG noted that EMA has reviewed dossiers presenting continuous manufacturing application (e.g., measure of humidity of the granules and on that basis the system adapting the process to ensure that at the end of the process the material was of acceptable quality). QIG noted this is less complex than the GSK digital twin but agreed that the same principles of performance-based controls can apply.

ASME V&V 40 and CDRH Credibility Guidance

- **Model credibility** refers to the trust in the predictive capability of the computational model for the COU.
 - **Question of interest** - describes the specific question, decision or concern that is being addressed.
 - **Context of use** - defines the specific role and scope of the computational model used to inform that decision.
 - **Model risk** - possibility that the model may lead to a false/incorrect conclusion about device performance, resulting in adverse outcomes.



Formally, ASME V&V 40 and CDRH guideline do not apply to data-based models. ASME VVUQ 70 sub-committee is developing standard for AI/ML model credibility.

Model Verification Proposal

Strategy for initial market supply

- Process boundaries will be defined based on the model and then experimentally verified
- The adaptive model will be active during manufacture to control the process within the design space
- The product control strategy will remain unchanged (CQAs controlled as part of batch release)
- Adaptive model performance can be demonstrated as part of PPQ where superiority of model-based control can be demonstrated over classical (fixed parametric) controls.
- The data from PPQ and subsequent CPV will show that the process is in a state of control
- **Model validation should not be required and only limited data on the model required in the dossier**

Increased model impact

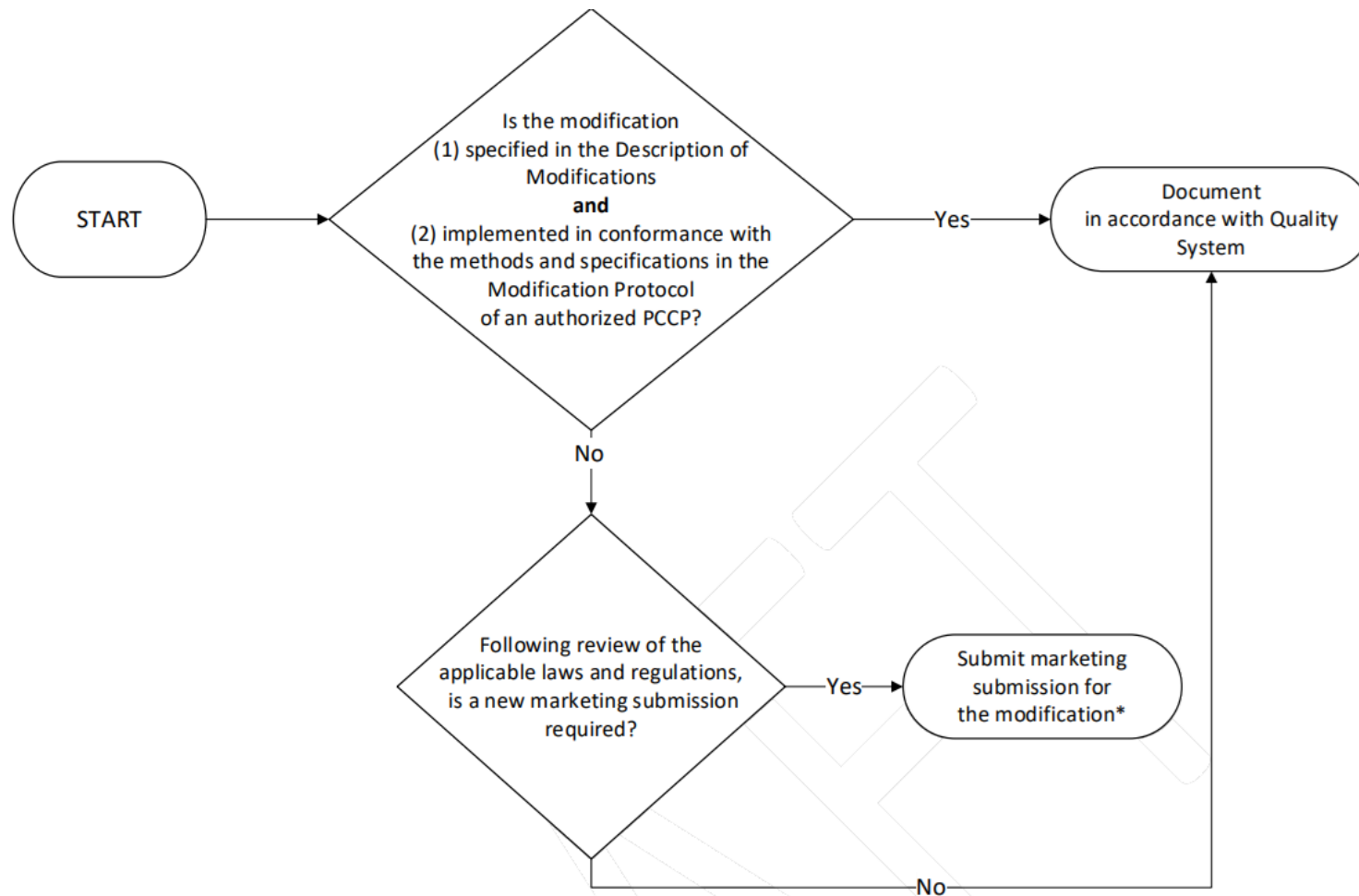
Future State

- Further model data provided in the dossier
- Flexible requirements for dossier content than can be defined/evolved via guidance
- **Reduced end-product testing**
- **Simplified PPQ**
- **Performance-based control strategy**

Framework will be required to ensure that changes to models can be managed under the site PQS without requiring prior approval

Predetermined Change Control Plans (PCCP)

- **PACMPs** give us the tool to use this approach for **high impact AI/ML** models.
- Use this approach to make **changes to performance-based ECs** for high-impact AI/ML models.



*For the modified device to have a PCCP, a PCCP should be submitted with the marketing submission so that the device and PCCP can be authorized together.

Other Things to Keep an Eye On: “Human In The Loop”

- **“For all models, especially those where there is no human-in-the-loop, a risk management plan should be developed** that defines likely risks of fail modes of the algorithm, e.g. what are the consequences of incorrect predictions/classifications as well as monitoring and mitigation/correction approaches, such as how to trigger a suspension/decommission of the model and how to suspend or decommission it.”
- Implication – according to this proposal, **by default**, an AI/ML model is higher risk than a human (e.g., do we establish a RMP for visual inspection?).
 - Is this really true, esp. for a GMP process with a well formulated, comprehensive control strategy in a GMP environment subject to PQS and routine inspections?

EMA Guidance on Process Models, Including AI/ML

1 22 February 2024
2 EMA/90634/2024
3
4

Comments should be provided using this EUSurvey [form](#). For any technical issues, please contact the [EUSurvey Support](#).

5 Preliminary QIG Considerations regarding Pharmaceutical 6 Process Models

7 Background

8 This Quality Innovation Group (QIG) document follows on from the first QIG Listen & Learn Focus
9 Group (LLFG) on Continuous manufacturing and the second QIG LLFG on Digital novel technologies,
10 held on 13 March 2023 and 12-13 October 2023 respectively. These highlighted the need for more
11 specific regulatory guidance on process models (hereafter called models).

12 It is recognised that regulatory expectations for process models in pharmaceutical manufacturing are
13 evolving; the intent of this document is to share QIG's current thinking with stakeholders and seek
14 their comments.

15 Introduction

16 Pharmaceutical process control consists of a series of measurements and actions within a process (or
17 system), designed to ensure that the desired quality of the output material is maintained over the
18 intended duration of process operation and over the lifecycle of a product. This includes measurements
19 and actions such as end point determinations, feed-forward/feed-back controls, statistical process
20 controls, and process monitoring.

21 Over the last few years, there has been an acceleration in the advancements for process control and
22 automation including sensor technology, data analytics and system modelling. The combination of
23 these innovative approaches creates a significant opportunity to enhance measurement and control of
24 process variables and output material attributes. This, in turn, supports adoption of advanced process
25 control strategies, continuous process verification, real-time process monitoring and optimisation, and
26 automated or even autonomous operation and management of manufacturing processes. Process
27 models play an increasingly important role in process design and validation, in control strategies and
28 during manufacturing process lifecycle. The expected outcome from the use of process models is
29 enhanced process understanding, (multivariate) monitoring and control, robustness, performance and
30 adaptability.

31 A model (in the context of pharmaceutical manufacturing) is a mathematical representation of a
32 physical or biological process or system. The model relates one or more input parameters to one or
33 more output parameters or properties relevant to the efficiency of the process and/or quality of the
34 material(s) being transformed by the system.

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- **Clarity on some very important issues:**

- Model impact vs. role in control strategy
- Emphasis on dossier content based on model performance
- Limited registration of algorithms
- Clarification of dossier content and validation requirements based on model impact

- **Thoughts:**

- Assessment of model risk in isolation?
- For low impact models, dossier content only necessary if model-based conclusions are filed?
- Clarity that all models would not need to strictly meet GMP
- Interesting section on “dual purpose” models predicting QAs as part of process design – need to think through this!
- Model lifecycle and maintenance protocol – important to clarify scope here as some models may not be maintained!

Considerations for the Use of Artificial Intelligence to Support Regulatory Decision-Making for Drug and Biological Products

Guidance for Industry and Other Interested Parties

DRAFT GUIDANCE

This guidance document is being distributed for comment purposes only.

<https://www.fda.gov/media/184830/download>

- **Initial Manufacturing Perspective:**
 - Guidance closely linked to ASME 40, which is supported by EMA and FDA and show the merging consensus on model-risk based approach to the deployment of AI.
 - Link to the control strategy and the QMS in mitigating risk is positive.
 - Potential to enable to development and deployment of AI in GMP manufacturing.
 - Example, Line 552 states: "In general, detailed plans for life cycle maintenance ((e.g. model performance metrics, risk-based frequency for monitoring...triggers for model retesting) **should be made available for review as a component of the manufacturing site's pharmaceutical quality system**, with a summary included in the marketing application for any product or process-specific models, in accordance with regulatory requirements"

FDA 21 CFR 211.110 Guideline

- Thoughts on this?

FDA is aware of industry's interest in using in-process control strategies that rely solely on process models to satisfy the requirements of § 211.110. This includes interest in strategies that use process models in continuous manufacturing to predict in-process material uniformity and homogeneity without any testing or examination of the in-process material (whether direct or indirect). However, to date, FDA has not been made aware of process models that demonstrate that: (1) the underlying assumptions of the process model will remain valid during routine manufacturing; and (2) the manufacturer can detect if an underlying assumption is no longer valid (e.g., a continuous mixing model that assumes uniform mixing would be unable to detect that uniform mixing is no longer occurring due to material agglomeration on the walls of the mixer). In other words, current process models cannot ensure the continued validity of all of the model's underlying assumptions at all times, particularly during certain unplanned disturbances. In the event of an unplanned disturbance that is not accounted for by the model's underlying assumptions, such control strategies would be unable to prevent nonconforming in-process materials (e.g., nonhomogeneous powder blend) from continuing through production and being used "in manufacturing or processing operations for which they are unsuitable."^{27,28} Therefore, control strategies that rely solely on current process models would be insufficient to satisfy the requirements of § 211.110.

ISPE AI CoP and RQHC Pharmaceutical Modeling Team

InTouch | November / December 2024

ISPE Announces ISPE AI®

ISPE recently announced ISPE AI®, an initiative aimed at aiding the pharmaceutical industry in realizing the potential of artificial intelligence (AI). The initiative will include a multifaceted approach to supporting the industry in AI readiness, beginning with the launch of the ISPE Community of Practice (CoP) on AI.

Over time, ISPE will also provide new ISPE Guidance Documents, additional conference sessions, new training courses, and more resources that focus on AI-related planning and implementation.

<https://ispe.org/pharmaceutical-engineering/november-december-2024/ispe-announces-ispe-air>

- **RQHC Pharmaceutical Modeling Team**
 - Feedback to EMA QIG Process Modeling Considerations Paper.

<https://ispe.org/sites/default/files/regulatory/2024/Comments%20from%20ISPE%20to%20EMA%20QIG%20for%20Process%20Models%20Postion%20FINAL.pdf>

GSK