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## ICH Quality IWG Points to consider for ICH Q8/Q9/Q10 guidelines

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# Points to consider for ICH Q8/Q9/Q10 guidelines

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# 1. Introduction

The ICH Quality Implementation Working Group (Q-IWG) has prepared 'Points to Consider' covering topics relevant to the implementation of ICH Q8(R2), Q9 and Q10, which supplement the existing [Questions & Answers](#) and [workshop training materials](#) already produced by this group. They should be considered all together.

The 'Points to Consider' are based on questions raised during the ICH Q-IWG training workshop sessions in the three regions. The Points to Consider are not intended to be new guidelines. They are intended to provide clarity to both industry and regulators and to facilitate the preparation, assessment and inspection related to applications filed for marketing authorizations.

The development approach should be adapted based on the complexity and specificity of product and process; therefore, applicants are encouraged to contact regulatory authorities regarding questions related to specific information to be included in their application.

Using the Quality by Design (QbD) approach does not change regional regulatory requirements but can provide opportunities for more flexible approaches to meet them. In all cases, GMP compliance is expected.

## 2. Criticality of quality attributes and process parameters

Scientific rationale and Quality Risk Management (QRM) processes are used to reach a conclusion on what are Critical Quality Attributes (CQAs) and Critical Process Parameters (CPPs) for a given product and process

The Quality Target Product Profile (QTPP) describes the design criteria for the product, and should therefore form the basis for development of the CQAs, CPPs, and Control Strategy.

The information developed to determine CQAs and CPPs will help to:

- Develop control strategy
- Ensure quality of the product throughout the product lifecycle
- Increase product and process knowledge
- Increase transparency and understanding for regulators and industry
- Evaluate changes

### 2.1. Considerations for establishing CQAs and CPPs

The introduction of ICH Q9 states that: *"...the protection of the patient by managing the risk to quality should be considered of prime importance"*. The QTPP provides an understanding of what will ensure the quality, safety and efficacy of a specific product for the patient and is a starting point for identifying the CQAs.

As part of risk assessment, risk analysis, as defined by ICH Q9 is: *'the qualitative or quantitative process of linking the likelihood of occurrence and **severity of harm**. In some risk management tools, the ability to detect the harm (detectability) also factors in the estimation of risk.'*

38 **Relationship between risk and criticality:**

- 39 • **Risk** includes severity of harm, probability of occurrence, and detectability, and therefore the level  
40 of risk can change as a result of risk management.
- 41 • **Quality attribute criticality** is primarily based upon severity of harm and does not change as a  
42 result of risk management.
- 43 • **Process parameter criticality** is linked to the parameter's effect on any critical quality attribute.  
44 It is based on the probability of occurrence and detectability and therefore can change as a result  
45 of risk management.

46 **Considerations for identifying and documenting CQAs can include the:**

- 47 • Severity of harm (safety and efficacy) before taking into account risk control and the rationale for  
48 distinguishing CQAs from other quality attributes.
- 49 • Link to the patient as described in the QTPP.
- 50 • Basis on which the CQAs have been developed (e.g. prior knowledge, scientific first principles, and  
51 experimentation).
- 52 • Inter-dependencies of the different CQAs.

53 **Considerations for identifying and documenting CPPs can include the:**

- 54 • Risk assessment and experimentation to establish the linkage between potential CPPs and CQAs
- 55 • Basis on which the CPPs have been identified (e.g. prior knowledge, scientific first principles, QRM,  
56 Design of Experiment (DoE), and other appropriate experimentation).
- 57 • Inter-dependencies of the different CPPs.
- 58 • Selected Control Strategy and the residual risk.

59 **CQAs and CPPs can evolve throughout the product lifecycle, for example:**

- 60 • Change of manufacturing process (e.g. change of synthetic route).
- 61 • Subsequent knowledge gained throughout the lifecycle (e.g. raw material variability,  
62 pharmacovigilance, clinical trial experience, and product complaints).

63 **2.2. Relationship of criticality to control strategy**

64 The identification and linkage of the CQAs and CPPs should be considered when designing the control  
65 strategy. A well-developed control strategy will reduce risk but does not change the criticality of  
66 attributes.

67 The control strategy plays a key role in ensuring that the CQAs are met, and hence that the QTPP is  
68 realised.

69 **3. Control strategy**

70 **3.1. Life-cycle of the control strategy**

71 The life-cycle of the control strategy is supported by Pharmaceutical Development, Quality Risk  
72 Management (QRM) and the Pharmaceutical Quality System (PQS) as described in the ICH guidelines  
73 ICH Q8, Q9, Q10.

74 The following points can be considered:

75 **Development of control strategy:**

- 76 • The control strategy is generally developed and initially implemented for production of clinical trial  
77 materials. It can be refined for use in commercial manufacture as new knowledge is gained.  
78 Changes could include acceptance criteria, analytical methodology, or the points of control (e.g.  
79 introduction of real-time release testing).
- 80 • Additional emphasis on process controls should be considered in cases where products cannot be  
81 well-characterized and/or quality attributes might not be readily measurable due to limitations of  
82 testing or detectability (e.g. microbial load/sterility).

83 **Continual improvement of the control strategy:**

- 84 • Consideration should be given to improving the control strategy over the life-cycle (e.g. in  
85 response to assessment of data trends over time and other knowledge gained).
- 86 • Continuous process verification is one approach that enables a company to monitor the process  
87 and make adjustments to the process and/or the control strategy, as appropriate.
- 88 • When multivariate prediction models are used, systems that maintain and update the models help  
89 to assure the continued suitability of the model within the control strategy.
- 90 • **Change management of the control strategy:**
- 91 • Attention should be given to outsourced activities to ensure all changes are communicated and  
92 managed.
- 93 • The regulatory action appropriate for different types of changes should be handled in accordance  
94 with the regional regulatory requirements.

95 **Different control strategies for the same product:**

- 96 • Different control strategies could be applied at different sites or when using different technologies  
97 for the same product at the same site.
- 98 • Differences might be due to equipment, facilities, systems, business requirements (e.g.  
99 confidentiality issues, vendor capabilities at outsourced manufacturers) or as a result of regulatory  
100 assessment / inspection outcomes.
- 101 • The applicant should consider the impact of the control strategy implemented on the residual risk  
102 and the batch release process.

103 **Knowledge management:**

- 104 • Knowledge management is an important factor in assuring the ongoing effectiveness of the control  
105 strategy.
- 106 • For contract manufacturing, knowledge transfer in both directions between the parties should be  
107 considered, particularly for model maintenance and/or updates, application of design space, and  
108 control strategies incorporating real-time release testing.

109

110

### 111 **3.2. Suitability of control strategy at different scales**

#### 112 **Management of risk on scale-up:**

- 113 • Risk associated with scale-up should be considered in Control Strategy development to maximize  
114 the probability of effectiveness at scale. The design and need for scale-up studies can depend on  
115 the development approach used and knowledge available.
- 116 • A risk based approach can be applied to the assessment of suitability of a Control Strategy across  
117 different scales. QRM tools can be used to guide these activities. This assessment might include  
118 risks from processing equipment, facility environmental controls, personnel capability, experiences  
119 with technologies, and historical experience (prior knowledge). See the ICH Q-IWG case study for  
120 examples.

#### 121 **Scale-up considerations for elements of Control Strategy:**

- 122 • *Complexity of product and process*
- 123 • *Differences in manufacturing equipment, facilities and/or sites*
- 124 • *Raw materials:*
  - 125 - Differences in raw material quality due to source or batch to batch
  - 126 variability
  - 127 - Impact of such differences on process controls and quality attributes
- 128 • *Process parameters:*
  - 129 - Confirmation or optimization
  - 130 - Confirmation of the design space(s), if used
- 131 • *In-process controls:*
  - 132 - Point of control
  - 133 - Optimization of control methods
  - 134 - Optimization and/or updating of models, if used
- 135 • *Product specification:*
  - 136 - Verification of the link to QTPP
  - 137 - Confirmation of specifications i.e. methods and acceptance criteria
  - 138 - Confirmation of RTRT, if used

### 139 **3.3. Specifications and certificate of analysis (CoA) for real-time release** 140 **testing (RTRT)**

141 The purpose of specifications and CoAs remains the same in the case of RTRT, but the way to develop  
142 them is different. RTR tests are considered to be specification testing methods and follow the  
143 established regional regulatory requirements for release specifications (as interpreted in e.g. ICH Q6A  
144 and ICH Q6B guidelines) together with other regional regulatory requirements (e.g. formats, GMP,  
145 batch acceptance decisions).

146 The use of RTRT has been addressed (see ICH Q8(R2) Section 2.5.; ICH Q-IWG Q&A Chapter 2.2). The  
147 following are points to be considered when developing a specification and CoA for RTRT:

#### 148 **Quality attributes:**

- 149 • Not all COAs need to be included in the specification.

150 • The attribute to be measured (e.g. surrogate for a CQA) can depend on the point of testing and/or  
151 control (e.g. materials, process steps, process parameters).

152 • Linking of the measured attribute to CQA and QTPP

153 **Methods of control:**

154 • The type of control used (e.g. models, PAT, test of isolated material, end product test, stability and  
155 regulatory test)

156 • Reference to the testing method used, if relevant

157 • Validation of control method

158 **Acceptance criteria:**

159 • Acceptance criteria at control point

160 • Criteria for stability and regulatory testing

161 **CoA elements:**

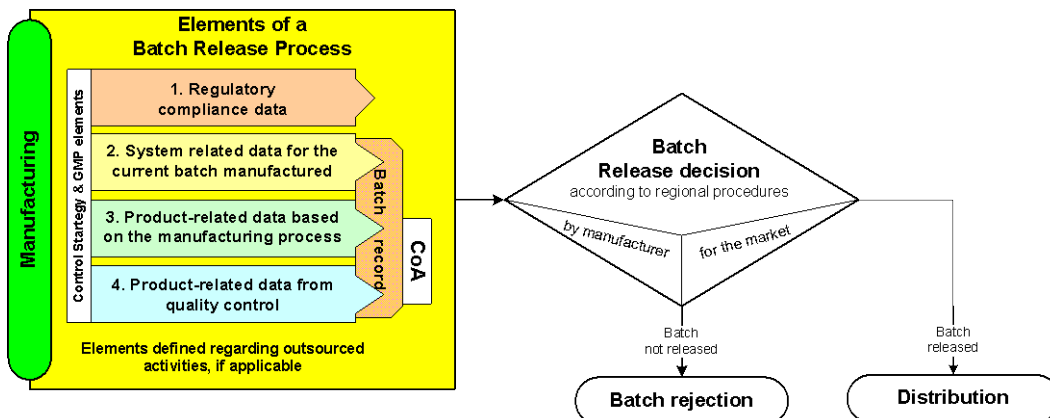
162 • Reported results e.g. values calculated from models, established calibrations and actual test results

163 • Acceptance criteria related to the method used

164 • Method references

165 **3.4. Process for a batch release decision**

166 Different development approaches lead to different control strategies. Regardless of the control  
167 strategy, the batch release process should be followed. For a batch release decision, several elements  
168 should be considered. See in the figure below an illustration of the elements of the batch release  
169 process leading to the batch release decision.



170

171 **1. Regulatory compliance data:**

172 There are regional differences in the regulation of batch release across the ICH regions [e.g.  
173 Qualified Person (EU), Good Quality Practice (Japan), Head of Quality Unit (US)] and the  
174 manufacturing licensing procedure. The PQS facilitates implementing and managing control  
175 strategy and Batch Release, notably through elements of a global approach (corporate / site /  
176 contractor). The PQS elements also facilitate regulatory compliance (e.g. changes that call for  
177 variation of the marketing authorization), including changes at manufacturing sites (e.g.  
178 changes regarding facilities, utilities and equipment).

179 **2. System related data for the current batch manufactured (e.g. environmental, facility,**  
180 **utilities and equipment)**

181 In the enhanced approach, there is an increased focus on process monitoring, which can  
182 provide the opportunity to perform continuous process verification. Any deviation or atypical  
183 event that occurs during manufacturing (e.g. involving the manufacturing process, facility,  
184 personnel, testing) is recorded and assessed, properly handled under the PQS (including CAPA)  
185 and closed out prior to release.

186 **3. Product-related data based on the manufacturing process**

187 Elements of the control strategy are defined and proposed in the marketing authorization  
188 dossier and agreed to by the regulators. Manufacturers should define, manage and monitor  
189 product-related data from batches manufactured according to the control strategy. These will  
190 be regularly assessed and reviewed during audits and inspections.

191 **4. Product-related data from quality control**

192 Results from end product testing and/or RTRT provide data based on which a CoA can be  
193 issued, in compliance with the specification as part of the release decision.

194 The **batch release process** leading to the **batch release decision** can be performed by more than  
195 one quality individual depending on the regional regulatory requirements and company policy:

- 196 • Batch release by manufacturer or contractor for internal purposes.  
197 • Batch release by manufacturing authorisation holder for the market.

198 **4. Level of documentation in enhanced (QbD) regulatory**  
199 **submissions**

200 This document is intended to provide suggestions on the type of information and the level of  
201 documentation that is appropriate to support a proposal for enhanced (QbD) approach. The type of  
202 information, as suggested in this document, is considered supportive and is intended to facilitate  
203 assessment and inspection without increasing the regulatory requirement. Submitted information  
204 should be organised in a clear manner and provide the regulators with sufficient understanding of the  
205 company's development approach; this information will be important to the evaluation of the proposed  
206 Control Strategy. Companies might consider, especially for QbD-containing submissions, an internal  
207 peer review process to assure quality, clarity and adequacy of the regulatory submission.

208 For submissions containing QbD elements (e.g. RTRT, design space), it is helpful for regulators to have  
209 a statement by the applicant describing the proposed regulatory outcome and expectations.

210 It is important to realize that not all the studies performed and/or data generated during product  
211 development need to be submitted. However, sufficient supporting information and data should be  
212 submitted in the application to address the following:

- 213 • The scientific justification of the proposed Control Strategy.  
214 • The scientific rationale for the studies conducted.  
215 • A concise description of methodologies used to conduct these studies and to analyze the generated  
216 data.  
217 • The summary of results and conclusions drawn from these studies.

218 The following sections include examples of background information that can be considered by both  
219 companies and regulatory authorities to assure scientific risk-based regulatory decisions.



## 220 **4.1. Risk management methodologies**

221 Following determination of the Quality Target Product Profile (QTPP) of the product under  
222 development, the applicant can use Quality Risk Management (QRM, ICH Q9) tools to rank and select  
223 quality attributes (including material attributes) and/or process parameters that should be further  
224 evaluated and/or controlled within appropriate ranges to ensure the desired product quality. The  
225 applicant should consider providing information of sufficient detail to demonstrate how the conclusions  
226 were reached, which can include:

- 227 • The scientific rationale for designation of QTPP and identification of corresponding CQAs (Critical  
228 Quality Attributes).
- 229 • Material attributes, process parameters and prior knowledge that were considered during risk  
230 assessment, preferably provided in a concise/tabulated form.
- 231 • Relevant known risk factors, e.g. degradation, solubility, etc.
- 232 • The scientific rationale and basis for the risk assessment as part of risk management and  
233 experiments that determined the final criticality of quality attributes and process parameters.
- 234 • Identification of potential residual risk that might remain after the implementation of the proposed  
235 Control Strategy (e.g. movements to commercially unverified areas of design space) and  
236 discussion of approaches for managing the residual risk.
- 237 • A list of critical and other quality attributes and process parameters.
- 238 • The linkage between CPP's, CQAs and the QTPP.
- 239 • Comment on the impact of the following on risk assessment: (a) interaction of attributes and  
240 process parameters, (b) effect of equipment and scale.

## 241 **4.2. Design of experiments**

242 The factors to be studied in a DoE could come from the risk assessment exercise or prior knowledge.  
243 Inclusion of a full statistical evaluation of the DoEs performed at early development stages (e.g.,  
244 screening) is not expected. A summary table of the factors and ranges studied and the conclusions  
245 reached will be helpful. For DoEs involving single- or multiple-unit operations that are used to establish  
246 CPPs and/or to define a Design Space (DS), the inclusion of the following information in the submission  
247 will greatly facilitate assessment by the regulators:

- 248 • Rationale for selection of DoE variables (including ranges) that would be chosen by risk  
249 assessment (e.g. consideration of the potential interactions with other variables).
- 250 • Any evidence of variability in raw materials (e.g. drug substance and/or excipients) that would  
251 have an impact on predictions made from DoE studies.
- 252 • Listing of the parameters that would be kept constant during the DoEs and their respective values,  
253 including comments on the impact of scale on these parameters.
- 254 • Type of experimental design used and a justification of its appropriateness, including the power of  
255 the design.
- 256 • Factors under study and their ranges can be presented in a tabular format. Submitters should  
257 indicate if the factors are expected to be scale-dependent.
- 258 • Reference to the type of analytical methods (e.g. HPLC, NIR) used for the evaluation of the data  
259 and their suitability for their intended use (e.g. specificity, detection limit).

- 260 • Results and statistical analysis of DoE data showing the statistical significance of the factors and  
261 their interactions, including predictions made from DoE studies relevant to scale and equipment  
262 differences.

### 263 **4.3. Manufacturing process description**

264 While preparing regulatory submissions, applicants should consider:

- 265 • Regional regulatory requirements with regard to the level of detail in describing manufacturing  
266 processes.
- 267 • Describing the proposed design space, including critical and other parameters studied, and its role  
268 in the development of the control strategy.

269 Manufacturing changes should be managed in accordance with regional regulatory requirements.

270 Where relevant, applicants can also consider submitting post-approval change management plans or  
271 protocols to manage post-approval manufacturing changes based on regional requirements.

## 272 **5. Role of Models in Quality by Design (QbD)**

273 A model is a simplified representation of a system using mathematical terms. Models can enhance  
274 scientific understanding and possibly predict the behaviour of a system under a set of conditions.

275 Mathematical models can be utilised at every stage of development and manufacturing. They can be  
276 derived from first principles reflecting physical laws (such as mass balance, energy balance, and heat  
277 transfer relations), or from data, or from a combination of the two. There are many types of models  
278 and the selected one will depend on the existing knowledge about the system, the data available and  
279 the. This document is intended to highlight some points to consider when developing and implementing  
280 mathematical models during pharmaceutical product development, manufacturing and throughout the  
281 product lifecycle. Other approaches not described in this document can also be used.

### 282 **5.1. Categorisation of Models**

283 Models can be categorised in multiple ways. The categorisation approaches used throughout this  
284 document are intended to facilitate the use of models across the lifecycle, including development,  
285 manufacturing, control, and regulatory processes.

286 For the purposes of regulatory submissions, an important factor to consider is the model's contribution  
287 in assuring the quality of the product. The level of oversight should be commensurate with the level of  
288 risk associated with the use of the specific model. The following is an example of such a categorisation:

289 I. *Low-Impact Models:*

290 These models are typically used to support product and/or process development (e.g.  
291 formulation optimisation).

292 II. *Medium-Impact Models:*

293 Such models can be useful in assuring quality of the product but are not the sole indicators of  
294 product quality (e.g. most design space models, many in-process controls).

295 III. *High-Impact Models:*

296 A model can be considered high impact if prediction from the model is a significant indicator of  
297 quality of the product (e.g. a chemometric model for product assay, a surrogate model for  
298 dissolution).

299 For the purpose of implementation, models can also be categorised on the basis of the intended  
300 outcome of the model. Within each of these categories, models can be further classified as low,  
301 medium or high, on the basis of their impact in assuring product quality.

302

303 Some examples of different categories based on intended use are:

304 • *Models for supporting process design:*

305 This category of models includes (but is not limited to) models for: formulation optimisation,  
306 process optimisation (e.g. reaction kinetics model), design space determination and scale-up.  
307 Models within this category can have different levels of impact. For example, a model for  
308 design space determination would generally be considered a medium impact model, while a  
309 model for formulation optimisation would be considered a low impact model.

310 • *Models for supporting analytical procedures:*

311 In general, this category includes empirical (i.e., chemometric) models based on data  
312 generated by various Process Analytical Technology (PAT)-based methods, for example a  
313 calibration model associated with a near infrared (NIR)-based method. Models for supporting  
314 analytical procedures can have various impacts depending on the use of the analytical method.  
315 For example, if the method is used for release testing, then the model will be high-impact.

316 • *Models for process monitoring and control:*

317 This category includes, but is not limited to:

318 • *Univariate Statistical Process Control (SPC) or Multivariate Statistical Process Control  
319 (MSPC)-based models:*

320 These models are used to detect special cause variability; the model is usually derived  
321 and the limits are determined using batches manufactured within the target conditions.  
322 If an MSPC model is used for continuous process verification along with a traditional  
323 method for release testing, then the MSPC model would likely be classified as a  
324 medium-impact model. However, if an MSPC model is used to support a surrogate for a  
325 traditional release testing method in an RTRT approach, then the model would likely be  
326 classified as a high-impact model.

327 • *Models used for process control (e.g. feed forward or feedback).*

328 Data-driven models should be developed through appropriately-designed experiments.  
329 These models are typically medium-impact or high-impact. For example, a feed  
330 forward model to adjust compression parameters on the basis of incoming material  
331 attributes could be classified as a medium-impact model.

## 332 **5.2. Developing and Implementing Models**

333 The following steps, if applicable, can be followed in a sequential manner, but, occasionally, it may be  
334 appropriate to repeat an earlier step, thus imparting an iterative nature to this process. The overall  
335 steps are:

- 336 1. Defining the purpose of the model
- 337 2. Deciding on the type of modeling approach (e.g. mechanistic or empirical) and the possible  
338 experimental/sampling methodology to be used to support the model development.
- 339 3. Selection of variables for the model; this is typically based on risk assessment, underlying  
340 physico-chemical phenomena, inherent process knowledge and prior experience.

341

- 342 4. Understanding the limitations of the model assumptions in order to:  
343 a) Correctly design any appropriate experiments,  
344 b) Interpret the model results, and  
345 c) Include appropriate risk-reduction strategies.
- 346 5. Collecting experimental data to support model development. These data can be collected at  
347 laboratory, pilot or commercial scale, depending on the nature of the model. It is important to  
348 ensure that variable ranges evaluated during model development are representative of  
349 conditions that would be expected during operation.
- 350 6. Developing model equations and estimating parameters, based on a scientific understanding of  
351 the process and collected experimental data.
- 352 7. Validating the model, as appropriate (see section 5.3).
- 353 8. In certain cases, evaluating the impact of uncertainty in model prediction on product quality  
354 and, if appropriate, defining an approach to reduce associated residual risk, e.g. by  
355 incorporating appropriate control strategies (this can apply to high-impact and medium-impact  
356 models).
- 357 9. Documenting the outcome of model development, including model assumptions, and  
358 developing plans for verification and update of the model throughout the lifecycle of the  
359 product. The level of documentation would be dependent on the impact of the model (see  
360 section 5.4).

### 361 **5.3. Model Validation and Model Verification during the lifecycle**

362 Model validation is an essential part of model development and implementation. Once a model is  
363 developed and implemented, verification continues throughout the lifecycle of the product.

364 The following elements can be considered for model validation and verification and are appropriate for  
365 high-impact models. In the case of well-established first principles-driven models, prior knowledge can  
366 be leveraged to support model validation and verification, if applicable. The applicability of the  
367 elements listed below for medium-impact or low-impact models can be considered on a case-by-case  
368 basis.

- 369 • **Setting acceptance criteria** for the model relevant to the purpose of the model and to its  
370 expected performance. In setting the acceptance criteria, variability in sampling procedure  
371 (e.g. for blending) could also be considered. In situations where the model is to be used to  
372 support a surrogate for a traditional release testing method, the accuracy of the model  
373 performance vs. the reference method could be considered. For example, a multivariate model  
374 (e.g. a Partial Least Squares (PLS) model), when appropriate, can be used as a surrogate for  
375 traditional dissolution testing. In this case, the PLS model is developed in terms of in-process  
376 parameters and material attributes and can be used to predict dissolution. One of the ways to  
377 validate and verify model performance in this case would be to compare accuracy of prediction  
378 of the PLS model with the reference method (e.g. a traditional dissolution method).
- 379 • **Comparison of the accuracy** of calibration vs. the accuracy of prediction. This can often be  
380 approached through internal cross-validation techniques using the same data as the calibration  
381 data set.
- 382 • **Validating the model** using an external data set (i.e., a data set from experiments/batches  
383 not used for model-building).

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- It can be beneficial to **verify the prediction accuracy** of the model by parallel testing with the reference method during the initial stage of model implementation and can be repeated throughout the lifecycle, as appropriate. If models are used to support a design space at commercial scale or are part of the control strategy, it is important to verify the model at commercial scale. For example, if a calibration model associated with a NIR-based method is developed at the laboratory scale and the method is then transferred to and used in commercial scale. Additionally, the data sets used for calibration, internal validation, and external validation should take into account the variability anticipated in future routine production (e.g. a change in the source of raw material that might impact NIR prediction). Low-impact models typically do not call for verification.

395 Approaches for model verification can be documented according to the pharmaceutical quality system  
396 (PQS) of the company and can include the following: a risk-based frequency of comparing the model's  
397 prediction with that of the reference method, triggers for model updates (e.g. due to changes in raw  
398 materials or equipment), procedures for handling model-predicted Out of Specification (OOS) results,  
399 periodic evaluations, and approaches to model recalibration.

#### 400 **5.4. Documentation of Model-related Information**

401 The level of detail for describing a model in a regulatory submission is dependent on the impact of its  
402 implementation in assuring the quality of the product. For the various types of models the applicant  
403 can consider including:

404 I. *Low-Impact Models:*

405 A discussion of how the models were used to make decisions during process development.

406 II. *Medium-Impact Models:*

407 Model assumptions, a tabular or graphical summary of model inputs and outputs, relevant  
408 model equations (e.g. for mechanistic models) either in the submission or via a reference,  
409 statistical analysis where appropriate, a comparison of model prediction with measured data,  
410 and a discussion of how the other elements in the control strategy help to mitigate uncertainty  
411 in the model, if appropriate.

412 III. *High-Impact Models:*

413 Data and/or prior knowledge (e.g. for established first principles-driven models) such as:  
414 model assumptions, appropriateness of the sample size, number and distribution of samples,  
415 data pre-treatment, justification for variable selection, model inputs and outputs, model  
416 equations, statistical analysis of data showing fit and prediction ability, rationale for setting of  
417 model acceptance criteria, model validation (internal and external), and a general discussion of  
418 approaches for model verification during the lifecycle.

## 419 **6. Design Space**

### 420 **6.1. Development of Design Space**

421 A design space can be updated over the lifecycle as additional knowledge is gained. Risk assessments,  
422 as part of the risk management process, help steer the focus of development studies and define the  
423 design space. Operating within the design space is part of the control strategy. The design space  
424 associated with the control strategy ensures that the manufacturing process produces product that  
425 meets the Quality Target Product Profile (QTPP) and Critical Quality Attributes (COAs).

426 Since design spaces are typically developed at small scale, an effective control strategy helps manage  
427 potential residual risk after development and implementation. When developing a design space for a  
428 single unit operation, the context of the overall manufacturing process can be considered, particularly  
429 immediate upstream and downstream steps that could interact with that unit operation-. Potential  
430 linkages to CQAs should be evaluated in design space development.

431 In developing design spaces for existing products, multivariate models can be used for retrospective  
432 evaluation of historical production data. The level of variability present in the historical data will  
433 influence the ability to develop a design space, and additional studies might be appropriate.

434 Design spaces can be based on scientific first principles and/or empirical models. An appropriate  
435 statistical design of experiments incorporates a level of confidence that applies to the entire design  
436 space, including the edges of an approved design space. However, when operating the process near  
437 the edges of the design space, the risk of excursions from the design space could be higher due to  
438 normal process variation (common cause variation). The control strategy helps manage residual risk  
439 associated with the chosen point of operation within the design space. When changes are made (e.g.  
440 process, equipment, raw material suppliers, etc.), results of risk review can provide information  
441 regarding additional studies and/or testing that might verify the continued applicability of the design  
442 space and associated manufacturing steps after the change.

443 Capturing development knowledge and understanding contributes to design space implementation and  
444 continual improvement. Different approaches can be considered when implementing a design space,  
445 e.g. process ranges, mathematical expressions, or feedback controls to adjust parameters during  
446 processing (see also Figure 1d in Q8(R2)). The chosen approach would be reflected in the control  
447 strategy to assure the inputs and process stay within the design space.

## 448 **6.2. 6.2. Verification and Scale-up of Design Space**

449 While the entire design space does not have to be re-established (e.g. DoE) at commercial scale,  
450 design spaces should be initially verified as suitable prior to commercial manufacturing. Design space  
451 verification should not be confused with process validation. However, it might be possible to conduct  
452 verification studies of the performance of the design space scale-dependent parameters as part of  
453 process validation. Design space verification includes monitoring or testing of CQAs that are influenced  
454 by scale-dependent parameters. Additional verification of a design space might be triggered by  
455 changes, e.g. site, scale, or equipment. Additional verification is typically guided by the results of risk  
456 assessments of the potential impacts of the change(s) on design space.

457 A risk-based approach can be applied to determine the design of any appropriate studies for  
458 assessment of the suitability of a design space across different scales. Prior knowledge and first  
459 principles, including simulation models and equipment scale-up factors, can be used to predict scale-  
460 independent parameters. Experimental studies could help verify these predictions.

## 461 **6.3. Documentation of Design Space**

462 Information on design space can be accommodated in the common technical document (CTD) in  
463 different presentation formats. Some examples of format and location in the document are covered in  
464 Q8(R2). Inclusion of a clear statement of the proposed design space and the location of the filed  
465 information (hyperlinked, where possible) in regulatory submissions should be considered to facilitate  
466 the regulatory process.

467 Some aspects of the design space that could be considered for inclusion in the regulatory submission:

468 The design space description including critical and other relevant parameters. The design space can be  
469 presented as ranges of material inputs and process parameters, graphical representations, or through  
470 more complex mathematical relationships

471 The relationship between the inputs (e.g. material attributes and/or process parameters) and the  
472 CQAs, including an understanding of the interactions among the variables.

473 Data supporting the design space, such as prior knowledge, conclusions from risk assessments as part  
474 of QRM and experimental studies with supporting data, design assumptions, data analysis, and models

475 The relationship between the proposed design space and other unit operations or process steps

476 Results and conclusions of the studies, if any, of a design space across different scales

477 Justification that the control strategy ensures that the manufacturing process is maintained within the  
478 boundaries defined by the design space

#### 479 **6.4. Lifecycle management of a Design Space**

480 The control strategy used for implementation of a design space in production depends on the  
481 capabilities of the manufacturing site. The batch records reflect the control strategy utilized. For  
482 example, if a mathematical expression is utilized for determining a process parameter or a CQA, the  
483 batch record would include the input values for variables and the calculated result.

484 As part of the technology transfer of a design space to a site and throughout the lifecycle, it is  
485 important to share the knowledge gained during development and implementation that is relevant for  
486 utilization of that design space both on the manufacturing floor and under the company / firm's/site's  
487 PQS. This knowledge can include results of risk assessments, assumptions based on prior knowledge,  
488 and statistical design considerations. Linkages among the design space, control strategy, CQA and  
489 QTPP are an important part of this shared knowledge.

490 Each company can decide on the approach used to capture design space information and movements  
491 within the design space under the company / firm's / site PQS, including additional data gained  
492 through manufacturing experience with the design space. In the case of changes to an approved  
493 design space, appropriate filings should be made to meet regional regulatory requirements. Movement  
494 within the approved design space, as defined in the ICH Q8(R2) glossary, does not call for a regulatory  
495 filing. For movement outside the design space, the use of risk assessment could be helpful in  
496 determining the impact of the change on quality, safety and efficacy and the appropriate regulatory  
497 filing strategy, in accordance with regional requirements.

### 498 **7. Process Validation / Continuous Process Verification**

499 These points to consider are intended to illustrate how using principles from ICH Q8, Q9 and Q10 can  
500 support an alternative Process Validation approach and are applicable to Drug Substance and Drug  
501 Product. They emphasise a more holistic approach to Process Validation across the product lifecycle,  
502 including Continuous Process Verification (CPV).

503 The main objective of Process Validation is to confirm that a process will consistently yield a product  
504 meeting its pre-defined quality criteria. This can be achieved in different ways, including a traditional  
505 approach, CPV, or a combination of these. There are different regional regulatory approaches to  
506 Process Validation. However, the concepts in this document are universally accepted, as is the  
507 appropriate use of Quality Risk Management principles in this context.

508

509 **7.1. General considerations**

510 In the traditional Process Validation approach the focus is on a limited number of batches at discrete  
511 time-points during the product lifecycle, e.g. at technology transfer or when changes are introduced.  
512 These batches are manufactured at commercial scale using the control strategy with an increased level  
513 and frequency of sampling. This validation approach remains appropriate, even if enhanced  
514 pharmaceutical development has been conducted.

515 Knowledge gained from development is the foundation for Process Validation. During technology  
516 transfer, site changes, and scale-up, the Control Strategy can be further developed as new variables  
517 are encountered in the commercial manufacturing environment. In many cases, new knowledge will be  
518 gained, often leading to modification of the Control Strategy and improvements to the process, thereby  
519 impacting Process Validation. This lifecycle approach to Process Validation recognises that elements of  
520 Process Validation begins with knowledge gained during development, and continues through  
521 technology transfer, and throughout the commercial manufacturing phase of a product.

522 A risk-based approach can be used to determine the plan for Process Validation studies, to ensure that  
523 process understanding is considered and that the areas of risk are addressed.

524 **7.2. Continuous Process Verification (CPV)**

525 ICH Q8 describes CPV as an approach to Process Validation that includes the continuous monitoring  
526 and evaluation of manufacturing process performance. Process Validation protocols can use CPV for the  
527 initial and on-going commercial production. CPV can also facilitate the evaluation of manufacturing  
528 process changes.

529 CPV can enhance the evaluation of the manufacturing process when it provides substantially more  
530 information on process variability and control.

531 CPV can be applied to an entire process, or to portions of a process, together with traditional Process  
532 Validation approaches.

533 Generally, for initial process validation, CPV is more appropriate when an enhanced development  
534 approach has been applied. However, it can also be used when extensive process knowledge has been  
535 gained through commercial manufacturing experience.

536 CPV can utilise in-line, on-line or at-line monitoring or controls to evaluate process performance. These  
537 are based on product and process knowledge and understanding. Monitoring can also be combined  
538 with feedback loops in order to adjust the process to maintain output quality. This capability also  
539 provides the advantage of enhanced assurance of intra-batch uniformity, fundamental to the objectives  
540 of Process Validation. Some process measurements and controls in support of Real Time Release  
541 Testing (RTRT) can also play a role in CPV.

542 Some advantages of CPV:

- 543 • Replaces the emphasis on the first few commercial-scale validation batches with enhanced  
544 assurance of product quality in many, or even all, batches
- 545 • Provides the foundation for a robust process performance and product quality monitoring  
546 system, increasing product and process knowledge and facilitation of continual improvement  
547 opportunities for process and product quality.
- 548 • Enables earlier detection of manufacturing-related problems and trends



- 549 • Provides immediate feedback of the effect of a change, thereby facilitating the management of  
550 changes.
- 551 • Provides a higher assurance of an ongoing state of control, as more data from CPV provide  
552 higher statistical confidence for ongoing monitoring and trending
- 553 • Is particularly suited to the evaluation of continuous manufacturing processes
- 554 • Contributes to the verification of the Design Space, if utilised, throughout the product lifecycle

### 555 **7.3. Pharmaceutical Quality System**

556 The Pharmaceutical Quality System (PQS) strengthens the link between the product lifecycle stages,  
557 thereby facilitating the Process Validation lifecycle approach. Data, information and knowledge from  
558 process performance and product quality monitoring, as described in ICH Q10, support the lifecycle  
559 validation approach and the continual improvement of the product and process.

560 Quality Risk Management, as an enabler for the PQS, contributes to process validation as follows:

- 561 • Risk assessment tools are useful in developing the Process Validation plan. This can also be  
562 useful for the evaluation of the effect of changes
- 563 • Statistical tools support monitoring and trending of process performance to assure a state of  
564 control.

565 Regardless of the approach to Process Validation, equipment and facilities should be suitably qualified,  
566 including computerised systems and control methods, as called for by GMP. Similarly, personnel  
567 involved in process validation activities should be appropriately trained and qualified.