

Maintenance of vaccine stability through annual stability and comparability studies

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WORKSHOP ON
STATISTICAL ANALYSIS OF STABILITY TESTING

21ST APRIL 2021

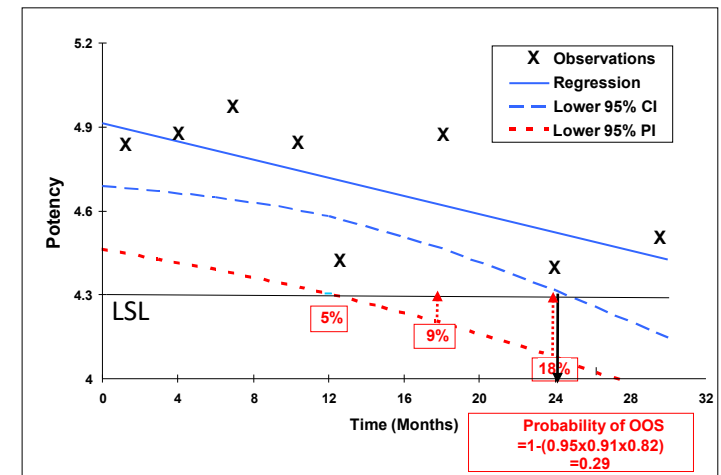


Opportunities for maintenance

- Stability monitoring
 - Annual stability program
- Stability comparability
 - When a process or product change may result in a change in kinetics
- Temperature excursions
 - Expected (LPI) and unexpected excursions in product storage conditions

Stability monitoring

- One lot per year to monitor product stability
- Stability OOS:
 - ICH Q1E advocates for use of a confidence interval to determine shelf life – represents the “average” of the stability profile
 - However, “individual stability measurements” are bounded by a much wider prediction interval
 - P(OOS) of individual measurements increases throughout shelf life
 - Culminating in ~30% chance of one or more OOS’s throughout shelf life



Schofield, *Maintenance*, 2009

Stability monitoring (cont.)

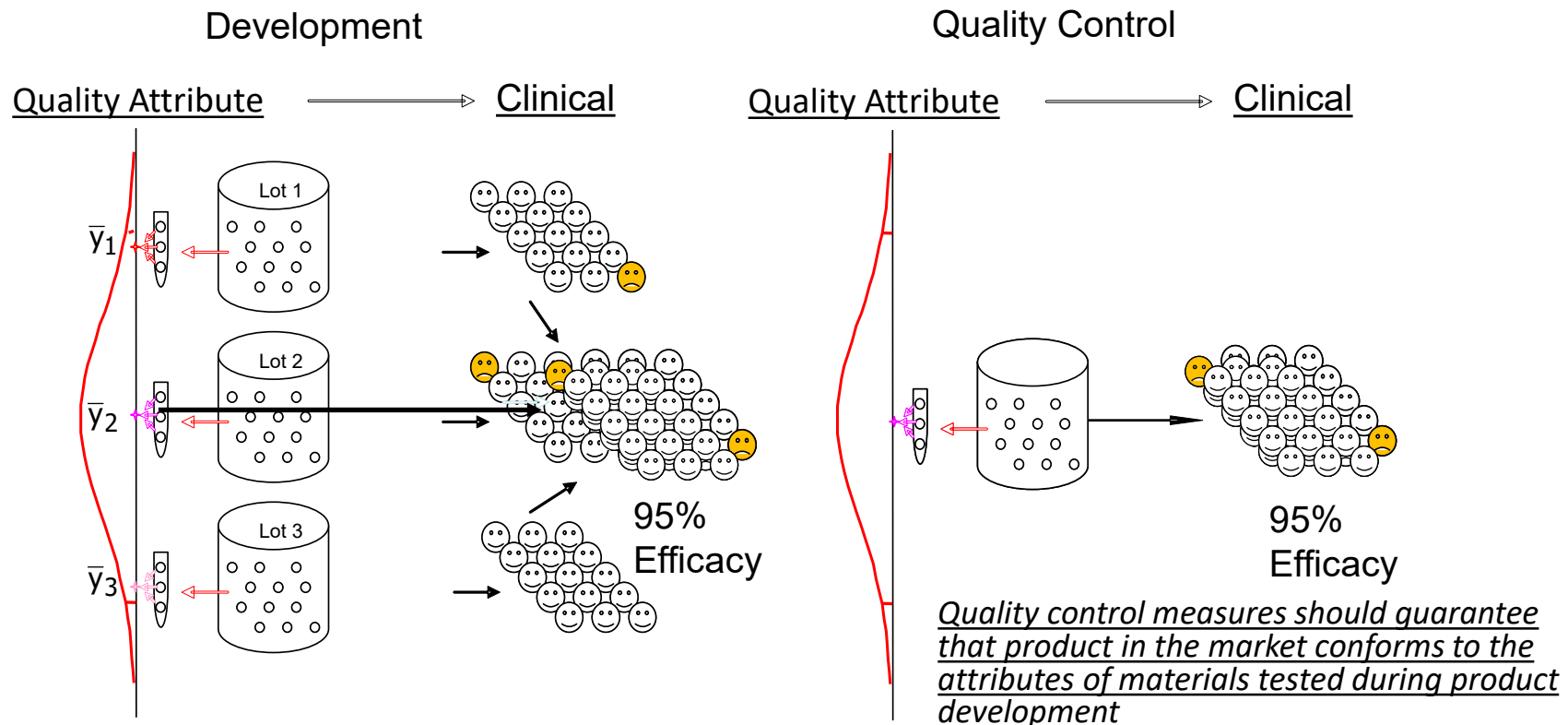
- What is the goal of stability monitoring?
 - Paradigm: Quality control measures should guarantee that product in the market conforms to the attributes of materials tested during product development
- What is the target population
 - What are we studying?



Individual Measurement
Individual Vial
Individual Time-Point
Individual Lot
Yearly Production
Product

Stability monitoring (cont.)

- Mean versus individual values
 - The mean of the batch is the measure of product quality

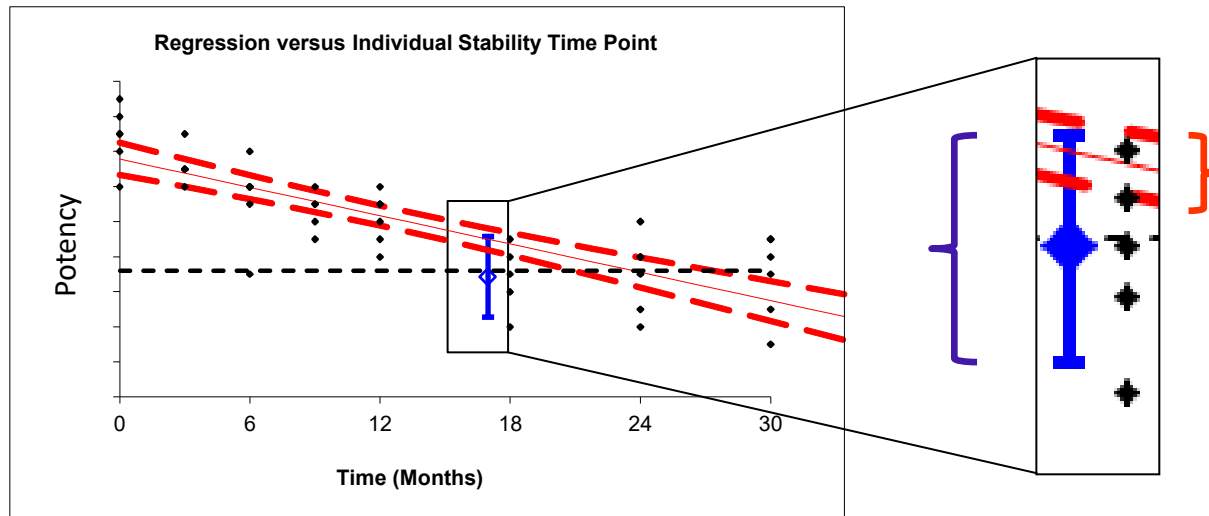


Stability monitoring (cont.)

- Solution: Treat post licensure stability as a form of “process monitoring”
 - Mitigation of stability OOS
 - Continued stability verification:
 - Use post licensure stability modeling
 - Combine data from ongoing post licensure studies to perform an overall analysis
 - Monitor slopes of post licensure studies

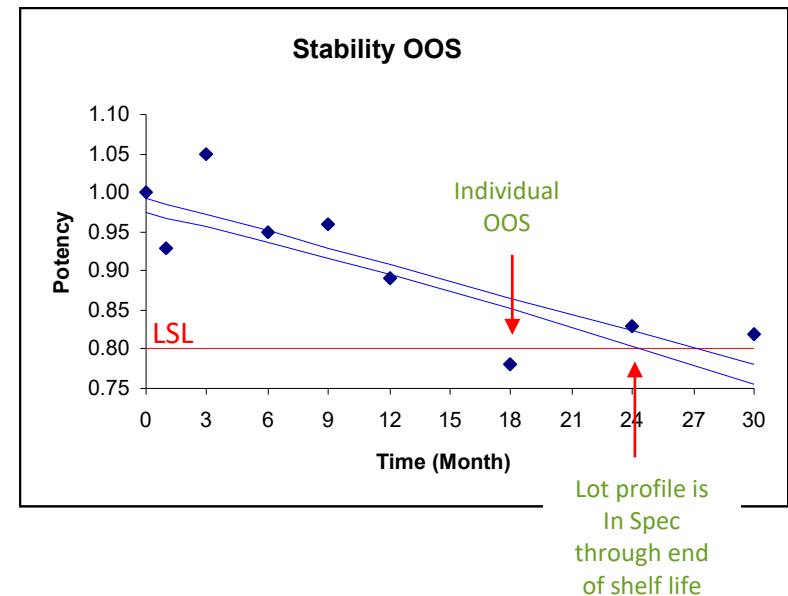
Stability monitoring (cont.)

- Estimate stability using an appropriate kinetics model
- Prediction from the kinetics model is more precise than prediction from individual stability measurements
 - “Smooths” out the long term variability of the potency assay
 - Uses the power of the measurements from other time points



Stability monitoring (cont.)

- Mitigation of stability OOS
 - Establish an OOT process using statistical modeling to demonstrate that the OOS result is not a quality concern, but is due to assay variability
 - Institute a retest plan to verify disposition of the lot
- Utilize the OOT process to predict quality and/or OOT



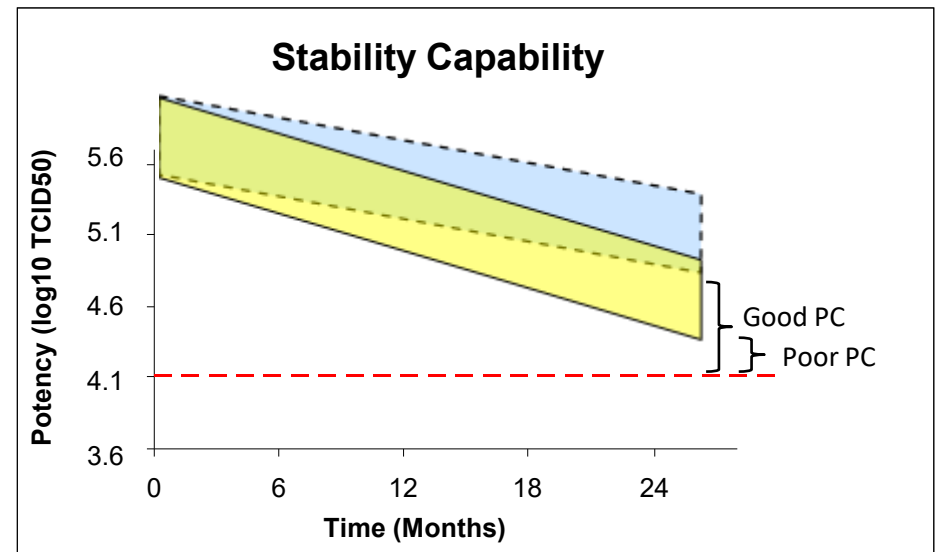
Gorko, 2003

Stability monitoring (cont.)

- Design a stability monitoring program
- Objective:
 - Utilize the post licensure program to bridge product stability performance to development, and to monitor the process for shifts and trends
- Design parameters:
 - Number of lots – Stability intervals – Assay format
- Design criteria:
 - Minimize the risk of missing a change in product stability (false success)
 - Minimize the risk of incorrectly detecting a change (false failure)

Stability monitoring (cont.)

- Select lots on a periodic basis (yearly, quarterly, monthly) based on “stability capability” - proximity of expiry potency to minimum potency
 - Good “SC” → less frequent selection
 - Poor “SC” → more frequent selection
 - Process monitoring:
 - Use combined data from ongoing stability lots to forecast expiry potency
 - Product monitoring:
 - Monitor individual lot slopes for extreme outliers.



Stability monitoring

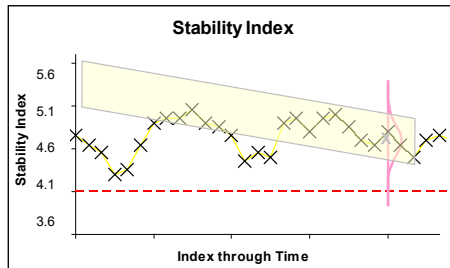
Process monitoring

- Combine data from ongoing post licensure studies to perform an overall analysis

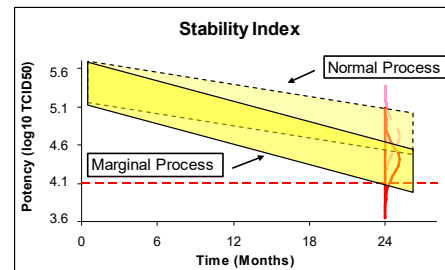
- Appropriate to a shelf life determination approach
- Formulated as an “Index”
- Equal to the predicted potency at EOSL from a pooled analysis

Sample calculation of vaccine quality index
4 Lots per Year

Number of months earlier released	Month							$w_i=1/\text{var}(Y)$	Predicted Month 24	$w_i \cdot \hat{Y}$		
0	4.891							NA	NA	NA		
3	4.771	4.932					0.009	6.06	0.0086			
6	4.932	4.723	4.727				0.040	4.08	0.0264			
9	4.926	4.556	4.613	4.934			0.115	4.81	0.0890			
12	4.933	5.012	4.334	5.028	5.862		0.263	6.16	0.2608			
15	4.846	4.975	4.887	4.245	4.011	3.821	0.528	3.10	0.2635			
18	4.811	5.098	5.105	4.576	4.983	5.042	5.271	0.966	5.19	0.8060		
21	4.934	4.769	4.545	4.605	4.244	4.770	3.501	5.774	1.647	4.63	1.2284	
24	4.894	4.632	4.351	4.668	4.388	5.105	4.586	4.627	3.222	2.647	4.11	1.7511
								Sum=	6.214		Index =	4.4338



Index is monitored over time with acquisition of each new time point



A threshold is determined which distinguishes a normal from a marginal process (predicts failure to meet LSL with 95% confidence).

Fairweather, 2003
Schofield, 2006

Stability monitoring

Process monitoring

Full		Month										$w_i=1/\text{var}(Y)$
Released	0	3	6	9	12	15	18	21	24			
0	4.891										NA	
3	4.771	4.932									0.009	
6	4.932	4.723	4.727								0.040	
9	4.926	4.556	4.613	4.934							0.115	
12	4.933	5.012	4.334	5.028	5.862						0.263	
15	4.846	4.975	4.887	4.245	4.011	3.821					0.528	
18	4.811	5.098	5.105	4.576	4.983	5.042	5.271				0.966	
21	4.934	4.769	4.545	4.605	4.244	4.770	3.501	5.774			1.647	
24	4.894	4.632	4.351	4.668	4.388	5.105	4.586	4.627	3.222		2.647	
									Sum=		6.214	

Drop 1		Month										$w_i=1/\text{var}(Y)$
Released	0	3	6	9	12	15	18	21	24			
0	4.891										NA	
3	4.771	4.932									0.009	
6	4.932	4.723	4.727								0.040	
9	4.926	4.613	4.934								0.112	
12	4.933	5.012	5.028	5.862							0.260	
15	4.846	4.975	4.887	4.011	3.821						0.494	
18	4.811	5.105	4.576	4.983	5.271						0.690	
21	4.934	4.769	4.605	4.244	4.770	5.774					1.220	
24	4.894	4.632	4.351	4.388	5.105	4.586	3.222				1.944	
							Sum=				4.768	

Drop 2		Month										$w_i=1/\text{var}(Y)$
Released	0	3	6	9	12	15	18	21	24			
0	4.891										NA	
3	4.771	4.932									0.009	
6	4.932	4.727									0.040	
9	4.926	4.934									0.101	
12	4.933	5.012	5.862								0.202	
15	4.846	4.887	4.011	3.821							0.472	
18	4.811	4.576	4.983	5.271							0.688	
21	4.934	4.769	4.244	5.774							1.391	
24	4.894	4.351	4.388	5.105	3.222						1.472	
					Sum=						4.375	

Testing Scheme	#Tests/Yr	%Saving	Index Sigma	%Decrease in Efficiency
Full	24	0%	0.4041	0%
Drop 1	20	17%	0.4580	12%
Drop 2	15	38%	0.4781	15%

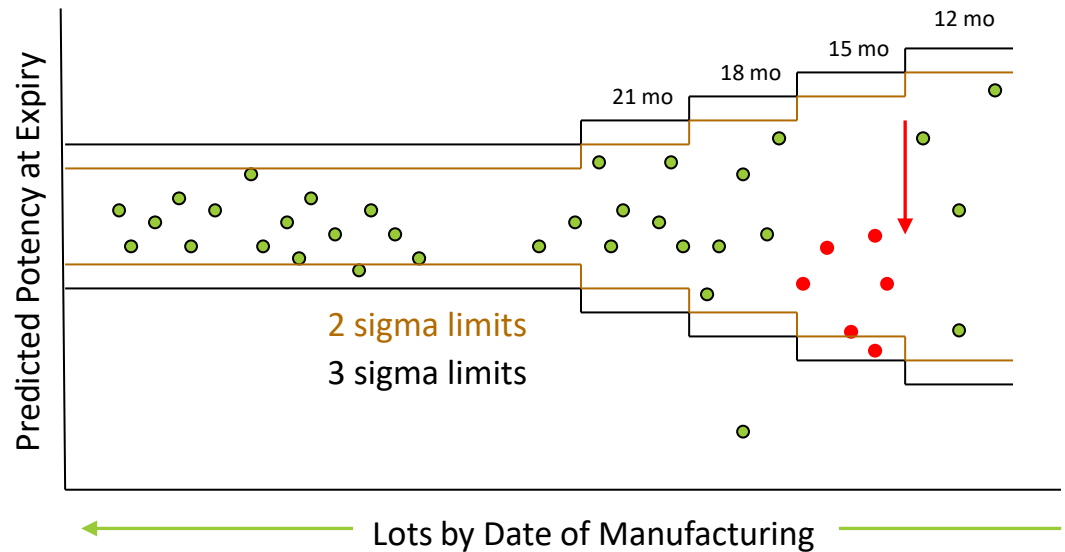
- Utilize “matrixing” to decrease stability testing burden
 - Dropping one test per lot in the first and second year results in a 17% saving in test burden, with a 12% drop in Index efficiency
 - Dropping two tests per lot in the first and second year results in a 38% saving in test burden, with a 15% drop in Index efficiency

Stability monitoring

Product monitoring

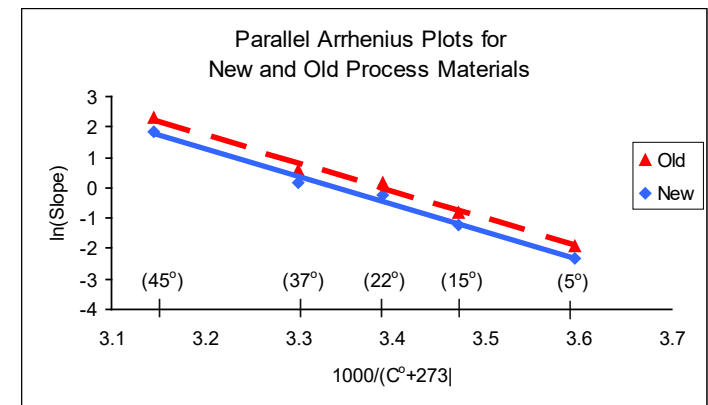
- Monitor slopes of post licensure studies (SPC)

- Appropriate to a release limit approach
- Evaluate ongoing lots with > 12 mos. of data. If predicted potency at expiry is outside 3 sigma limits, lot is investigated as an extreme outlier (atypical lot)
- Address potential shifts in a timely manner



Stability comparability

- Using Arrhenius relationship as a stability “fingerprint”
 - Not as a predictor of slope at the labeled storage temperature
 - With a commitment to monitor routine stability on new process materials
 - Statistical approaches have been proposed using a “stability space” and “equivalence testing”
 - ... or rely upon continuous stability verification (maybe together with accelerated stability)



$$\ln|\text{Slope}| = a + b \cdot \frac{1}{|\text{Temp}|}$$

Schofield, *Maintenance*, 2009
Noël, 2001
Burdick, 2011, 2013
Yu, 2015

Temperature excursions

- Excursions from the labeled storage condition occur for planned and unplanned reasons
 - Planned excursions under the manufacturers control including labeling, packaging, and inspection
 - Managed through a release model
 - Unplanned excursions outside the manufacturers control including equipment failures (e.g., refrigerators), regional practices (e.g., pharmacy to patient to doctor), and ECTC

Temperature excursions (cont.)

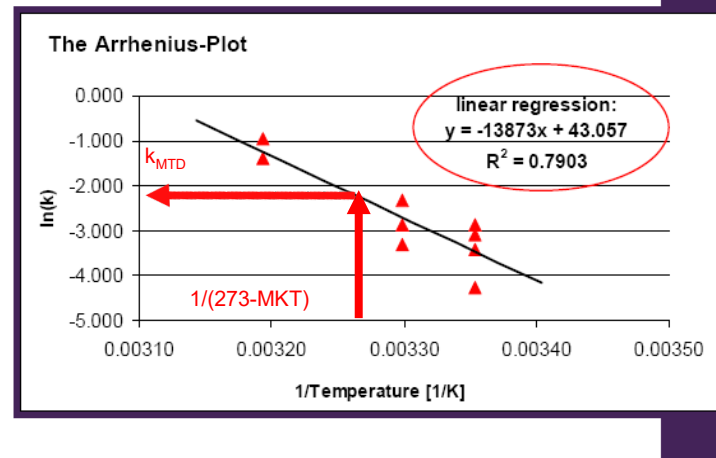
- An assessment plan can be developed using accelerated stability data and the Arrhenius model:

$$\ln(k) = a + \frac{b}{^{\circ}K}$$

- Determine mean kinetic temperature of excursion in $^{\circ}K$ ($273 - \text{MKT}_{^{\circ}C}$)
- Interpolate the degradation rate corresponding to mean kinetic temperature from the model – k_{MKT}
- Determine loss of potency due to excursion:

$$\text{Loss} = k_{\text{MKT}} \cdot t_{\text{Excursion}}$$

Example: Product AZ



MKT is expressed as:

$$-\ln \left(\frac{e^{-\Delta H/RT_1} + e^{-\Delta H/RT_2} + \dots + e^{-\Delta H/RT_n}}{n} \right)$$

Where.

ΔH = activation energy (typically from 60 to 100 kJ/mol for solids and liquids)

R = 8.314472 J/mol-K (universal gas constant)

T = temperature in degrees K

n = the number of sample periods over which data is collected

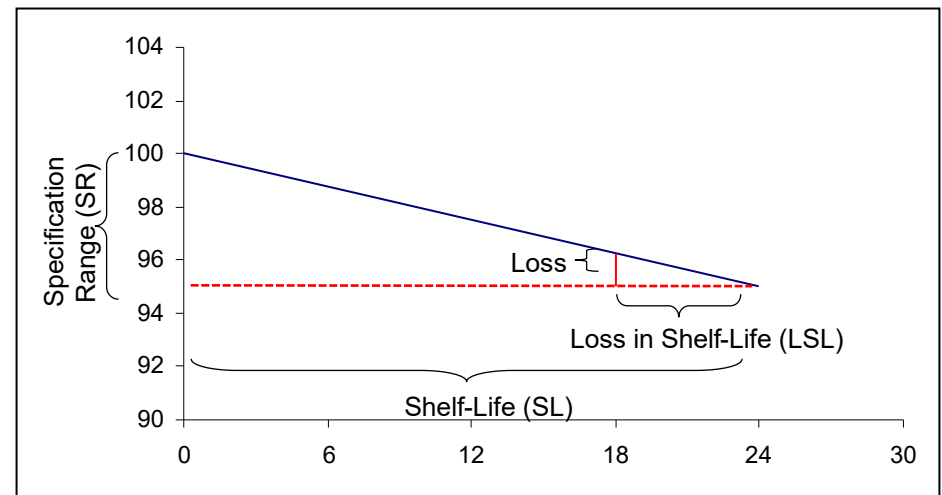
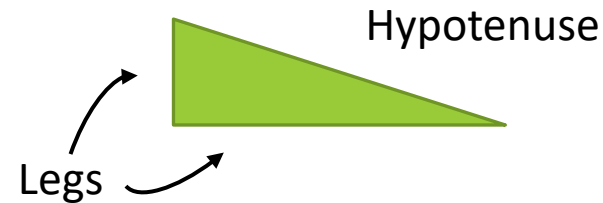
Note : \ln is the natural log and e is the natural log base.

Temperature excursions (cont.)

- Recalculate expiration date
- Use $Loss = k_{MKT} \cdot t_{Excursion}$ to determine the loss in shelf life (LSL) – amount of shelf life lost due to the excursion
- Using the principle of similar triangles – ratio of “legs” of similar triangles are equal

$$LSL / Loss = SL / SR,$$

$$LSL = SL / SR \cdot Loss$$



Temperature excursions (cont.)

- Example: a refrigerated product is subject to MKT exposure of 25°C over a 48-hour period of time.
 - Loss rate at 25°C = 0.0025 from the Arrhenius interpolation
 - Specification Range = Release – LSL = 3.5 – 3.0 = 0.5 log
 - Shelf-life = 24-Months

$$Loss = 0.0025 \log/hr \cdot 48 - hrs = 0.12 \log,$$

$$LSL = \frac{24 \text{ Months}}{0.5 \log} \cdot 0.12 \log = 5.8 \text{ Months}$$

- Thus, if expiry is 1/1/20, the recalculated expiry is ~7/1/19

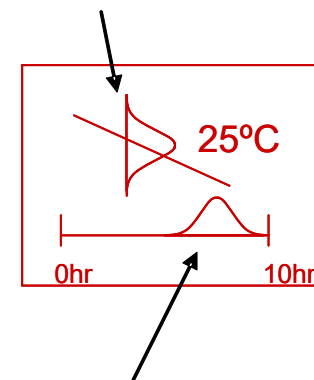
Temperature excursions (cont.)

- Assess the risk to a patient of receiving product that is unsafe or subpotent due to exposure to elevated temperatures outside the chain of custody of the manufacturer (e.g., pharmacy to patient to doctor)
 - A release model is built upon “worst case” exposure (i.e., maximum prescribed time of exposure such as product shelf life) to various conditions
 - A risk analysis simulates “real case” outcomes from models and information on the actual product exposures

Temperature excursions (cont.)

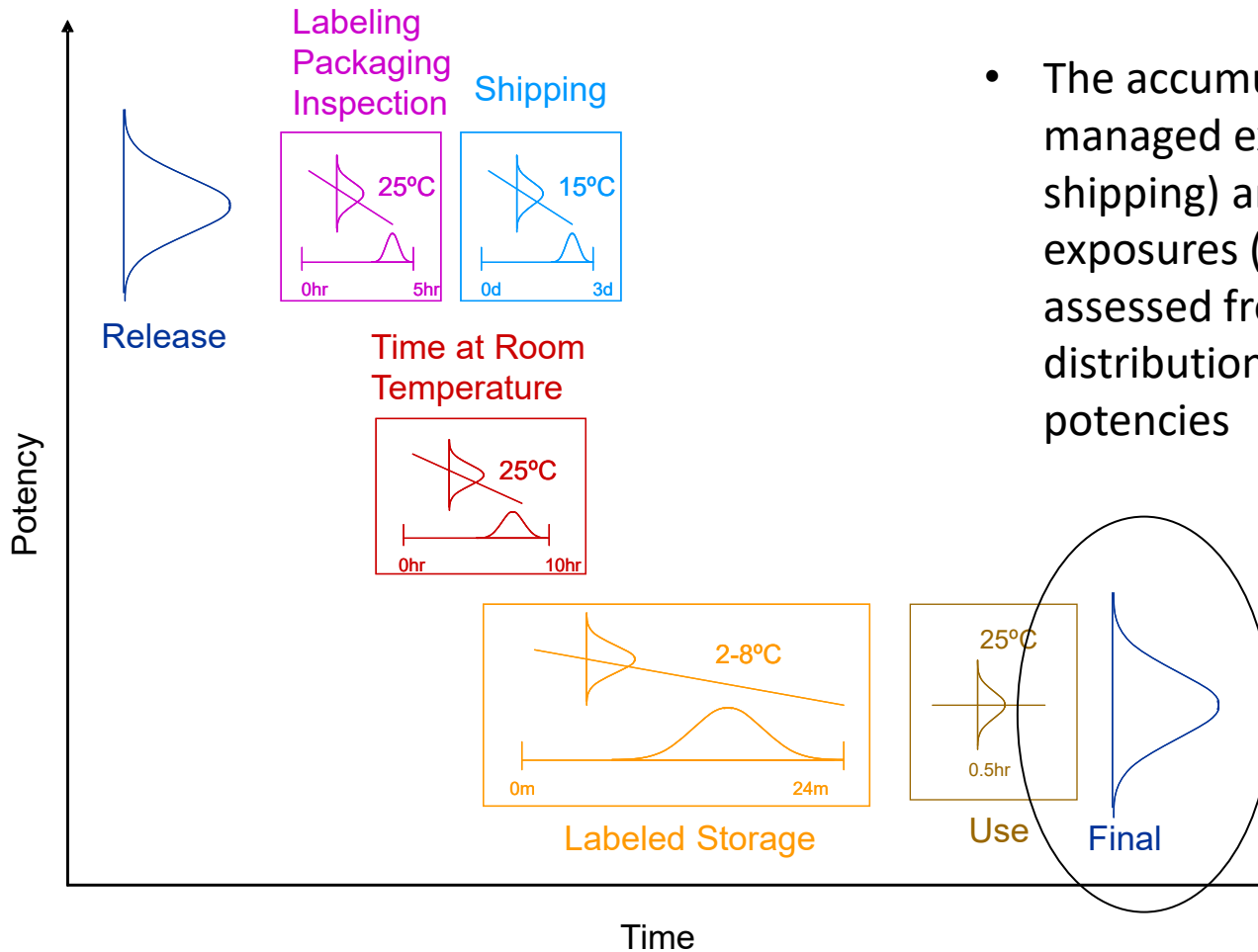
- Example of Monte Carlo simulation of expected potencies for material exposed to 25°C up to 10-hours
 - Simulate 10K random lots
 - Randomly pick from distributions of loss rates (b_i) and exposure times (t_j)
 - For each lot calculate the potency at the end of their exposure times
 - $Final = Beginning - b_i \cdot t_j$

Average loss rate and uncertainty (b_i)



Average time and variability of exposure (t_j)

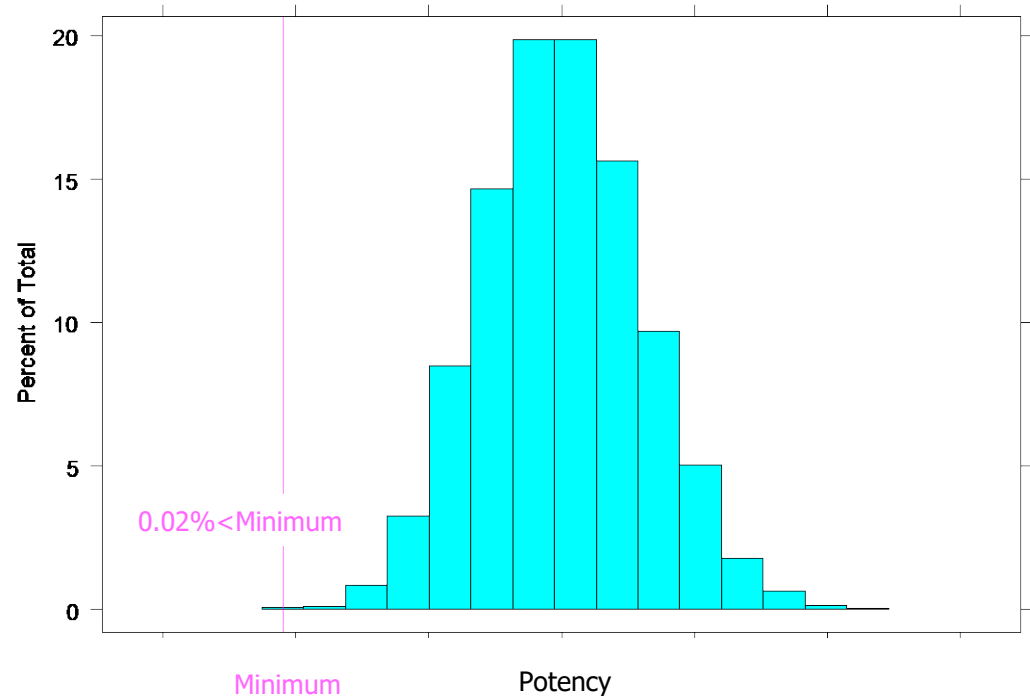
Temperature excursions (cont.)



- The accumulated impacts of managed exposures (LPI and shipping) and unmanaged exposures (time at RT) can be assessed from the simulated distribution of final (expiry) potencies

Temperature excursions (cont.)

- The resulting distribution is used to calculate percent of lots that are predicted to fall below the minimum potency specification at the time of administration
- The conclusion is that there is minimal risk (0.02%) that a patient will receive vaccine from a lot with potency less than the minimum specification
 - Note: the risk that a patient will receive subpotent vaccine is 5% using the release model



Summary

- Product quality is assured through appropriate calculation of shelf-life and release limit
- Stability monitoring can be designed and analyzed to ensure that the “stability process” and “product stability” deliver potent vaccine through the end of shelf life
- Stability comparability can effectively use accelerated temperature conditions to address relevant process changes
- Intended excursions can be managed using the release model; unintended excursions can be managed by reassessing the expiry date from the estimated excursion loss

References

1. WHO *Guidelines for Stability Evaluation of Vaccines* (2006)
2. WHO *Guidelines on the stability evaluation of vaccines for use under extended controlled temperature conditions* (ECTC, 2015)
3. Schofield, TL (2009) *Vaccine stability study design and analysis to support product licensure*; *Biologicals* 37 (2009) 387-396.
4. Schofield, TL (2009) *Maintenance of vaccine stability through annual stability and comparability studies*; *Biologicals* 37 (2009) 397-402
5. Fairweather WR, Mogg R, Bennett PS, Zhong J, Morrissey C, Schofield TL. (2003) *Monitoring the stability of human vaccines*. *Journal of Biopharmaceutical Statistics*; 13: 395-413.
6. Schofield, TL, et.al. (2006) *Monitoring the stability of human vaccines*, presented at WCBP, SF
7. Gorko, MA (2003) *Identification of Out-of-Trend Stability Results*, *Pharmaceutical Technology*; 27(4)
8. Noël C, Charles S, Francon A, Flandrois JP (2001) *A mathematical model describing the thermal virus inactivation*. *Vaccine*;19:3575-82.
9. Yu, B, Zeng, L (2015) *Evaluating the comparability of stability at long-term storage temperature using accelerated stability data*, IABS Statistical Meeting, September 29-30
10. Sidor, L, Burdick, R, Cowley, D, Kendrick, BS, (2011) *Demonstrating comparability of stability profiles using statistical equivalence testing*, *BioPharm International*; 24 ,36-42
11. Burdick, RK, Sidor, L, (2013) *Establishment of an equivalence acceptance criterion for accelerated stability studies*, *Journal of Biopharmaceutical Statistics*; 23, 730-743

Thank you

Questions?

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