

# Process Performance Analysis for Roche's Pharmaceutical Manufacturing Network

**Yiming Peng, Theo Koulis, Jens Lamerz, and Dan Coleman** Nonclinical Biostatistics Genentech, A Member of the Roche Group

**2017 ASA Biopharmaceutical Section Nonclinical Biostatistics Conference** Rutgers University, New Jersey Jun 12 - 14, 2017



#### Contents



- 1. Motivation and strategy
- 2. Roche's process performance index, Rpk
- 3. Process Performance Analysis (PPA) pilot study
- 4. FAQ
- 5. PPA roll out

#### **Motivation**

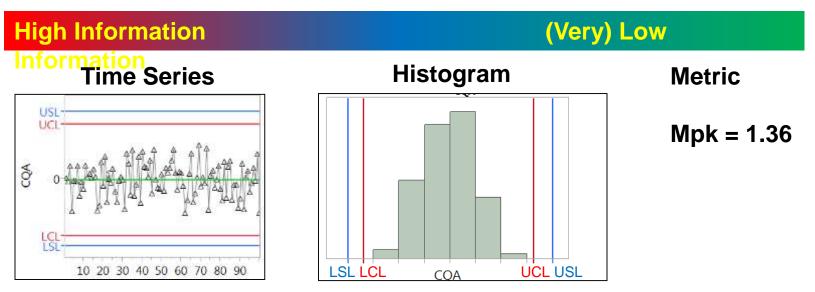


Objective	Characterizing our global manufacturing network and identifying opportunities for continuous process improvement through risk ranking of Critical Quality Attributes (CQA)
Client	Senior management and Health Authorities
Scope	<ul> <li>CQA from large and small molecules, drug substance (DS) and drug product (DP)</li> <li>~30 Products</li> <li>~2 Sites per Product</li> <li>~20 CQAs per Product</li> <li>~20-200 batches per CQA per site per year</li> </ul>

#### How to summarize 1,200 time series per year?

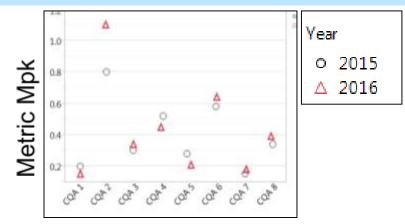
#### **Summarizing Time Series**



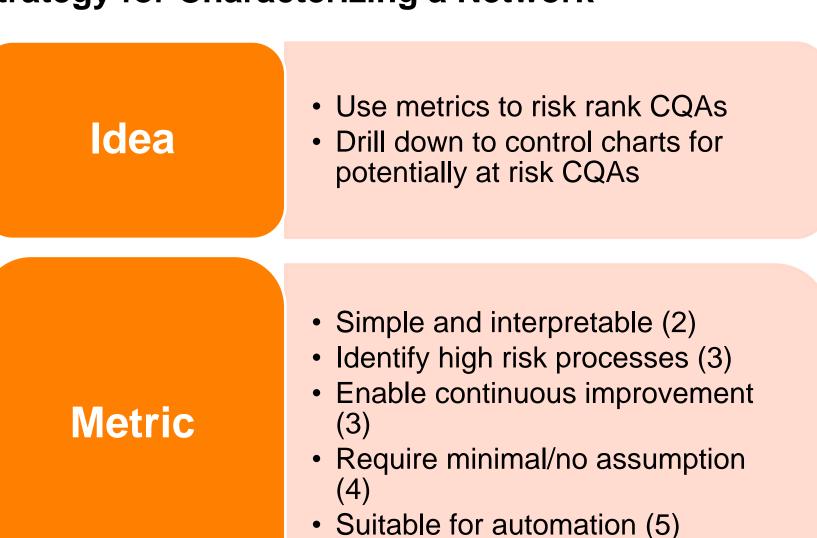


#### Metrics a necessary evil?

#### **Metrics Enable Comparisons of Time Series**



#### **Strategy for Characterizing a Network**





# 2. Roche's process performance index, Rpk

# Simple and Interpretable

### **Risk index Rpk**

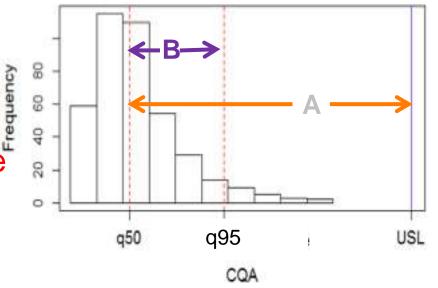


- R<sub>pk</sub> consistent description of process performance that does not depend on distribution (Clement (1989))
- R<sub>pk</sub> proportion of the allowable range used by the process
  - $-R_{pk}$  is typically between 0 and 1
  - 0 indicates high process performance
  - 1 indicates low process performance
  - <u>Small is Beautiful, Big is Bad!</u>
- Use (k=u, I, 2)
  - $-R_{pu}$  for attributes with upper specific. limits, e.g., impurities
  - $-R_{pl}$  is for attributes with lower spec. limits, e.g., monomer
  - $-R_{p2}$  is for attributes with upper and lower spec. limits, e.g., potency <sub>7</sub>

# R<sub>pu</sub> definition (u = upper specification limit)

- B = <u>Bulk of data</u>, Percentage of the data between q50 and q95
- A = <u>Allowable Range</u>, interval between q50 and USL
- <u>R<sub>pu</sub></u> = the proportion of allowable range used by the bulk of the data
- $R_{pu} = B/A$





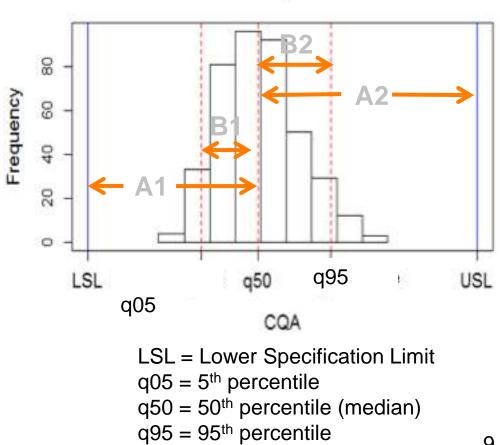
 $q50 = 50^{th}$  percentile (median)  $q95 = 95^{th}$  percentile USL = Upper Specification Limit





# $R_{p2}$ definition (2 = two-sided specification)

- $R_{p2} = max\{B1/A1,$ B2/A2}
- R<sub>p2</sub> is the largest proportion of allowable range used by a bulk of the data



USL = Upper Specification Limit

Histogram

9



# 3. Pilot study results

# Identify high risk processes

# Enable continuous improvement

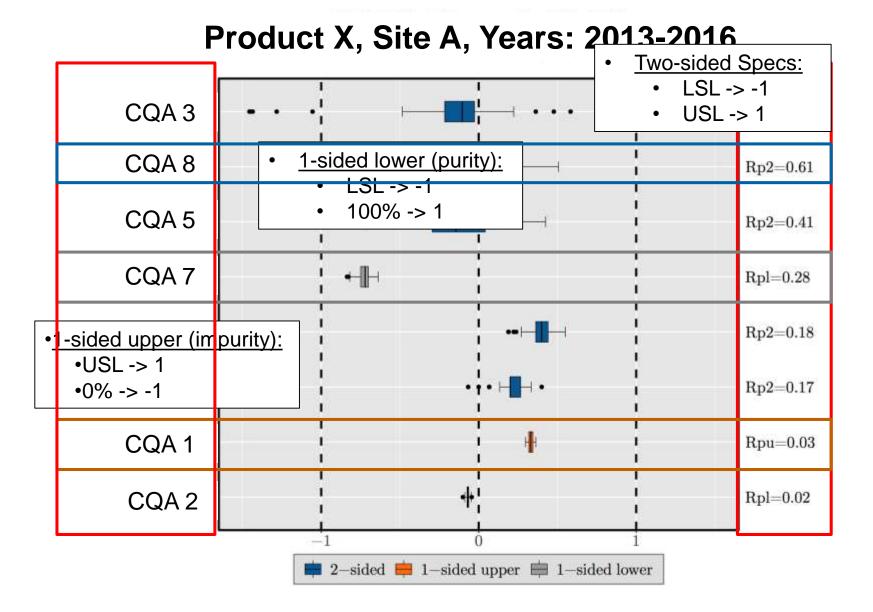


# **Pilot Study**

- Data:
  - -9 products
    - 5 Biologics, 4 Small Molecule
  - DS and DP
  - Multiple years
  - Multiple sites
- Analysis
  - PPA at site level (single product, single site)
  - PPA at product level (single product, multi-site)
  - PPA at network level (multi-product, multi-site)

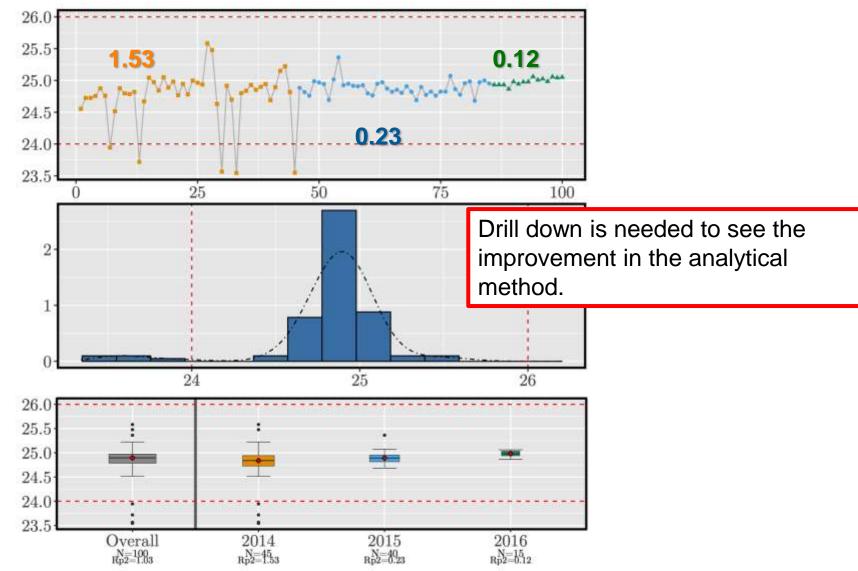


#### Site level: Risk-Ranking Overview



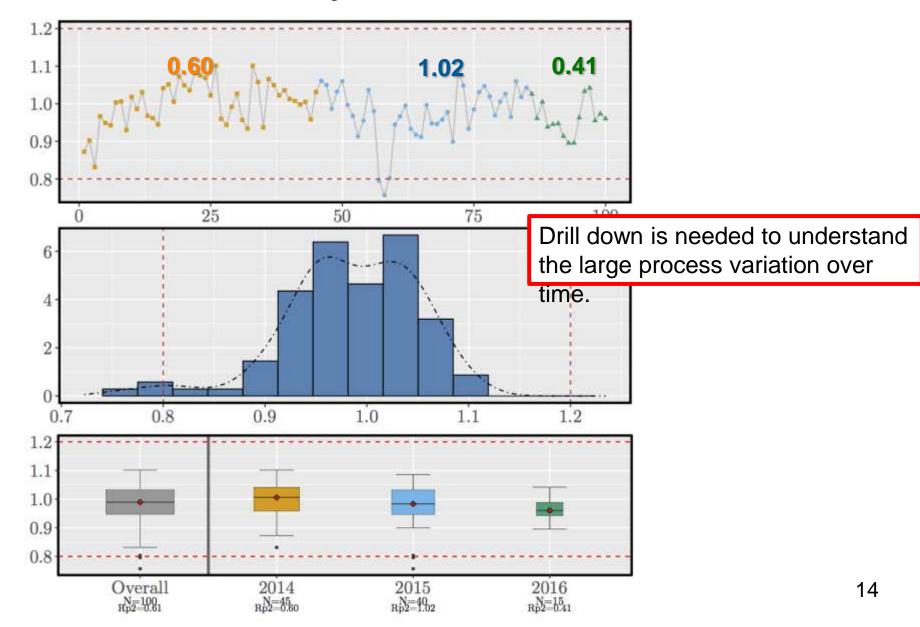


#### Drill Down: CQA 3 by Year





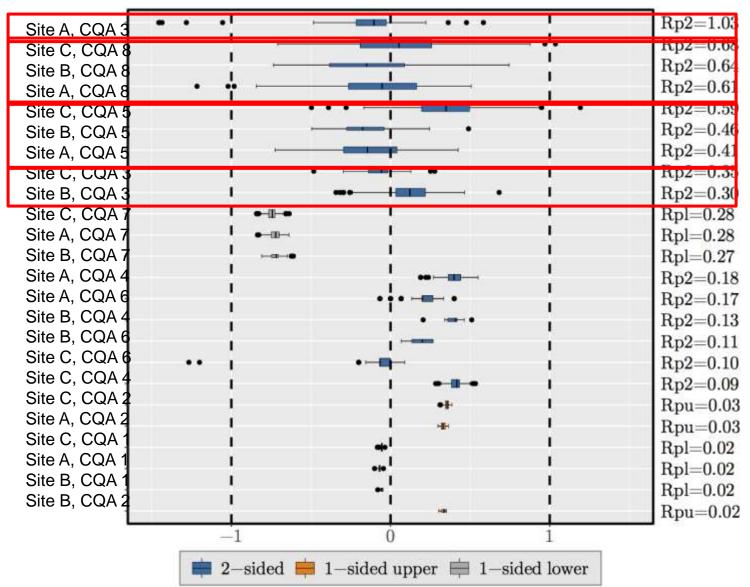
#### **Drill Down: CQA 8 by Year**





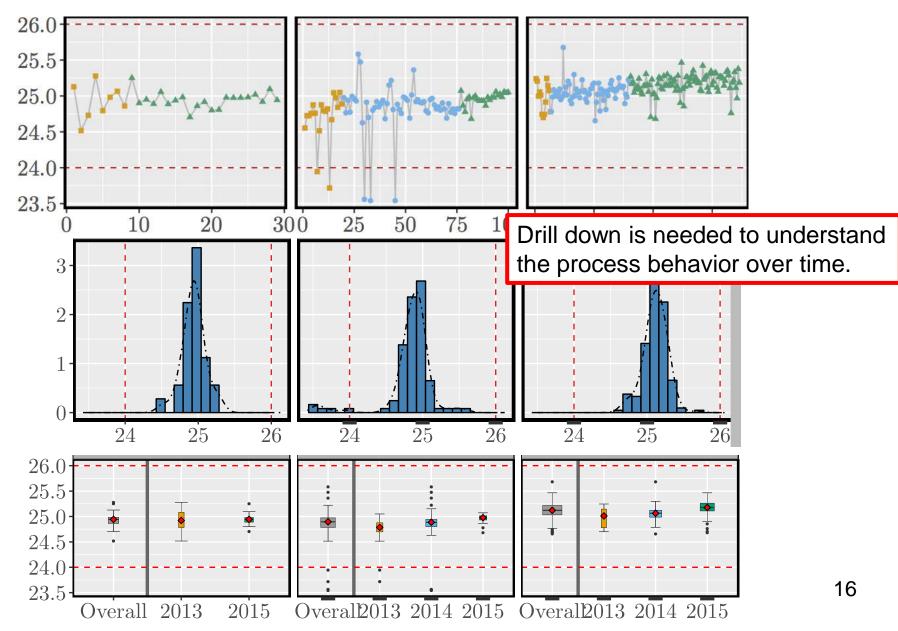
#### **Product level: risk ranking overview**

#### Product X, Site A B C, Years: 2013-2016



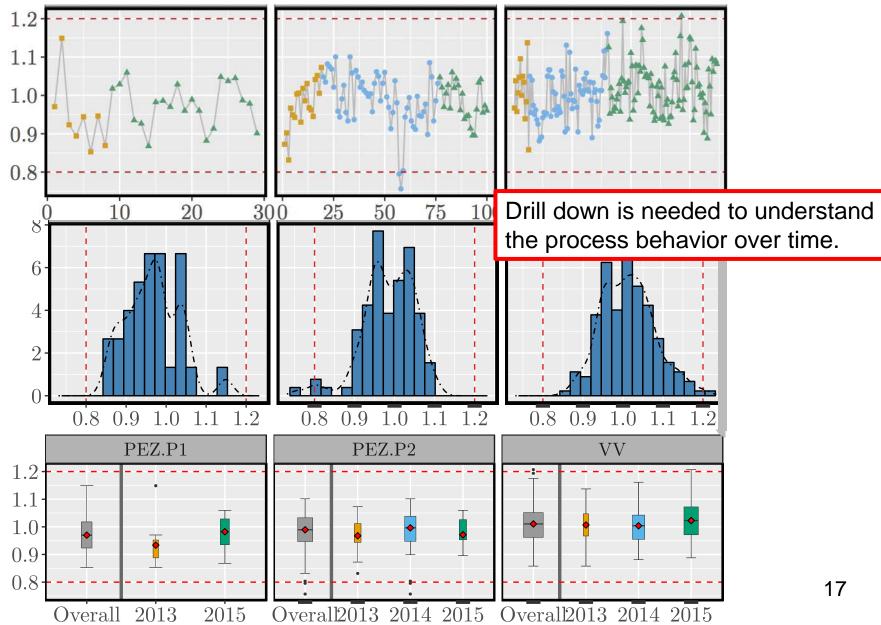
#### Roche

#### Drill Down: CQA 3 by Year, Multi-site



#### Roche

#### Drill Down: CQA 8 by Year, Multi-site



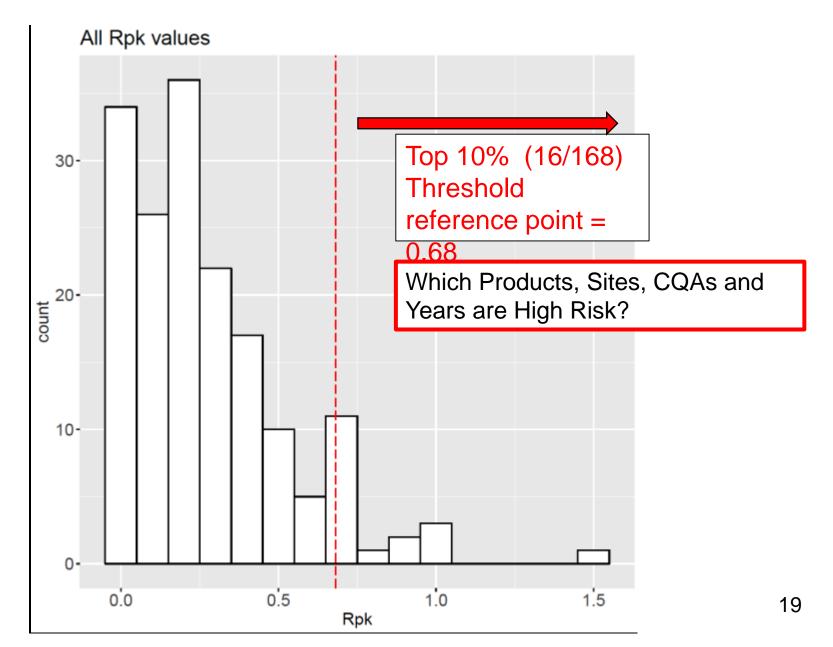
#### **Network level:**



- Review network manufacturing process performance
- Design threshold to drive continuous process improvement

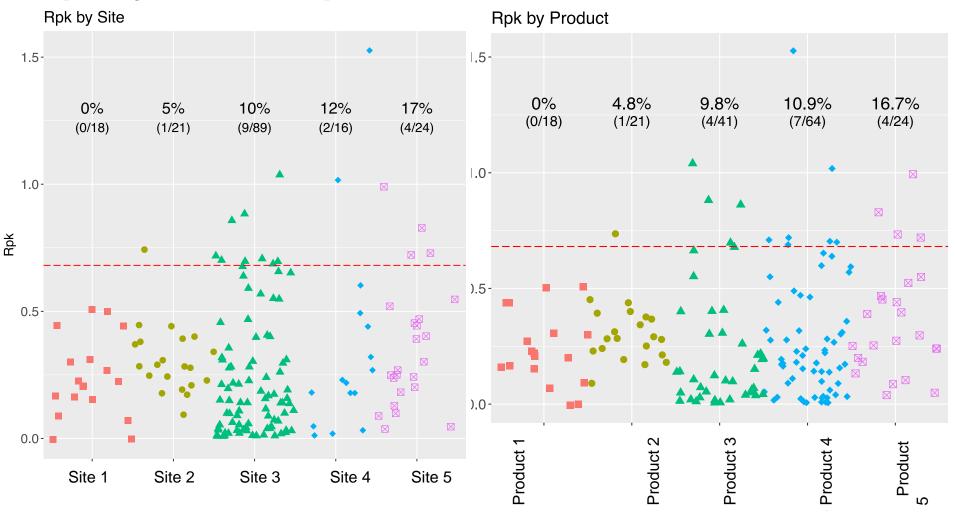
### Annual Rpk for Biologics (N≥20 batches)





#### Roche

#### Characterization of Biologics Network Rpk by sites and product



20

#### **Empirically derived Thresholds**



#### Request:

Manufacturing sites want thresholds to classify Rpk as low or high risk

#### Proposal:

Use 10% percentile from network results, from previous year to set threshold for the next year.

Possibly separate thresholds for biologics DS, biologics DP, small molecule DS, and small molecule DP.

#### **Properties:**

If process performance does not change, expect 10% high risk and 90% low risk

If process performance improves, expect fewer high and more low risk



### 4. FAQ

# Rpk requires minimal / no assumption

#### Q1: Why not Cpk or Ppk?

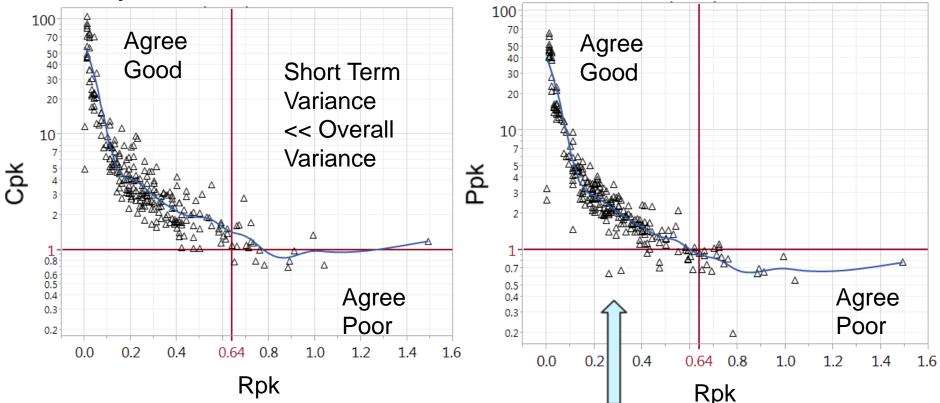


- Cpk uses the **short term variation**, which excludes variation due to trends, shifts, and other sorts of process wandering
- These components of the overall variation need to be included for risk assessment -- Cpk is not suitable for risk ranking

- Ppk uses the overall variation, but similar as Cpk, the calculation assumes process data follows Normal distribution -- mean  $\pm$  3 sd describes the data when the distribution is Normal (symmetric)
- Many Pharmaceutical data may not follow Normal distribution Ppk and Cpk may be misleading



# Comparison of annual Rpk to Ppk and Cpk on pilot study (biologics & small molecules, N≥20)



Short tail distributions or special cause outliers inflating variance



# Q2: Why use the .05 and .95 quantiles, *q.05* and q.95? -- Why not the range, i.e., minimum and maximum, or the .00135 and .99865 **guantiles?** The range is not acceptable because interpreting it correctly and

- consistently depends on the sample size.
  - E.g. the range of 10 CQA values is likely to be a lot smaller than the range of 100 CQA values -- Quantiles are better
- q.05 and q.95 is a **practical compromise** between making the "bulk" as large as possible and having sufficient data to reliably compute the sample quantiles. For datasets with at least 20 = 1/.05 batches, our experience is that Rpk has been reliable for risk ranking
- *q..00135* and *q.99865* make the "bulk" of the data very large but require large sample sizes 741=1/0.00135 to estimate. For many pharmaceuticals, 741 batches are far more than would be made in a year



# Q3: Why is Rpk "the smaller the better" instead of "the larger the better"?

- Rpk behaves linearly with the process range, i.e., reduce the process range by half, the Rpk is reduced by half; double the process range, the Rpk is doubled – linearity makes it more interpretable
- When the process range is within specifications, Rpk is between 0 and 1, while the Cpk and Ppk are between 1 and infinity.



# Q4: Why do we not require processes to be in "statistical control" for Rpk to be valid?

- Conventional use of Cpk, e.g. in consumer products industries, demands that processes be in "statistical control" [Montgomery, 2009] so that the indices describe "stable distributions".
- This enables the data to meet strong statistical assumptions needed to predict future product results. Such demands are generally unrealistic for the pharmaceutical industry
- Rpk is designed for risk ranking by upper level managers: the aim is to provide an informative, accurate overview of network performance without making inappropriate statistical assumptions and oversimplifications.
- Rpk is meant to be **descriptive**, not predictive.



# Q5: What distributional assumptions must be met for Rpk to give results suitable for risk ranking?

• None

- Several "rules of thumb" to help assure the meaningful characterization
  - There should be at least 20 data points per CQA.
  - There should be at least 10 distinct values in the data



# **5. PPA Roll out**

# **Automation**

#### Suitable for automation



- No outlier detection, no data transformation, etc.
- R Shiny app (excel data in, word document report out)



#### Conclusions

- <u>Rpk</u>
  - Rpk is a reasonable metric for risk ranking CQAs
  - appropriate for all data, including low volume, nonnormal data
- Business drivers:
  - Rpk provides a common metric for characterizing a pharmaceutical manufacturing network
  - Goal is to make Rpk a component in management's decision making, e.g., prioritizing process improvement through risk ranking CQAs
- <u>Challenge:</u>
  - integration into the quality system



#### Acknowledgements

• Bert Gunter, Pia Krieger, David Cate, etc.

#### References

- "The Use and Abuse of Cpk," B.H. Gunter, Quality Progress, Parts 1-4 (1989-1991)
- Process Capability Indices Samuel Kotz and Norman L. Johnson
- Beyond Capability Confusion 2<sup>nd</sup> Ed. Donald J. Wheeler
- Advanced Topics in Statistical Process Control 2<sup>nd</sup> Ed. Donald J. Wheeler
- Statistical Quality Control 7<sup>th</sup> Ed. Douglas C. Montgomery
- "Statistical Quantiles in Statistical Packages" Rob J. Hyndman and Yanan Fan, The American Statistician (1996)
- "Process Capability Computations for Non-Normal Distributions" John A. Clement, Quality Progress (1989)
- "Process Capability Indices for One-sided specification intervals and skewed distributions – Kerstin Vannman and Malin Albing, Quality and Reliability Engineering International (2007)



# Doing now what patients need next