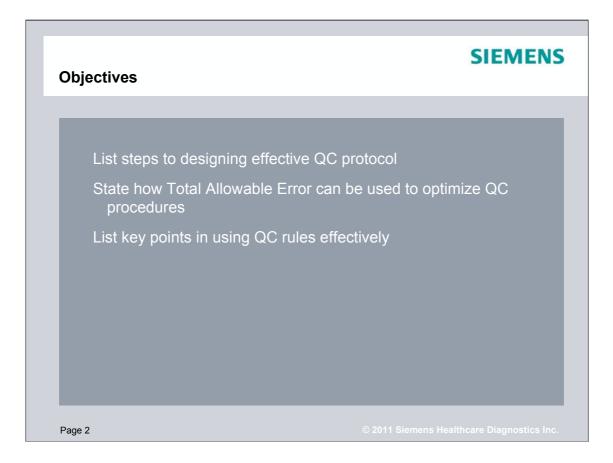
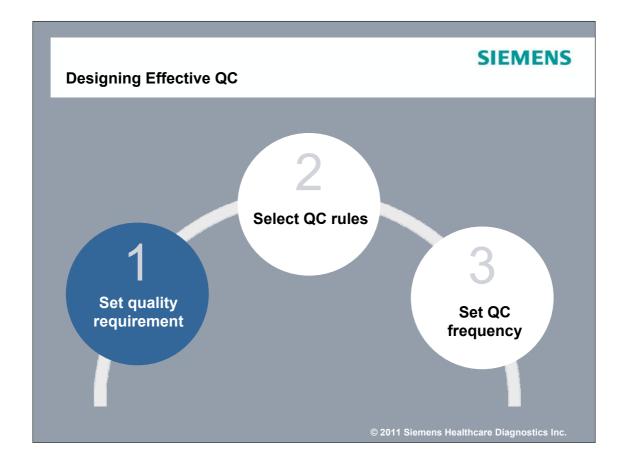


Welcome to the Siemens Healthcare Educational session on Designing Effective QC. For the next 90 minutes we are going to look at some of the tools available to design a laboratory QC protocol that can reliably detect significant change in method performance while also being cost effective and practical. Probably more cost effective and practical that what is commonly done in many laboratories today.

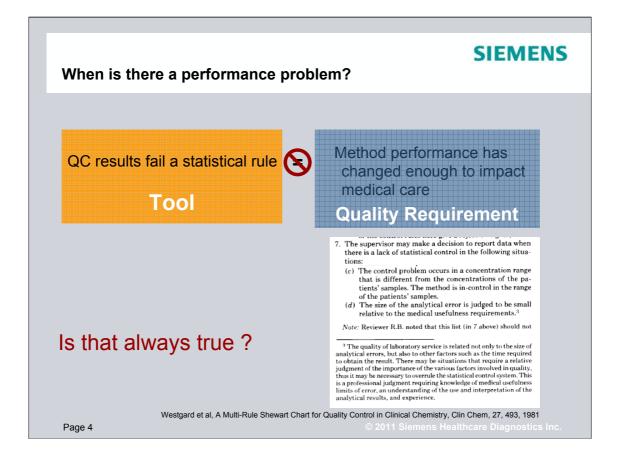


Our objectives for this session are to look at the steps involved in designing effective QC, to look at how Total Allowable Error can be effectively used in the process and to review some of the key points to keep in mind when using QC rules to enhance effectiveness.



There are three steps to designing effective QC. First set the quality requirement. Next Select the QC rules that will be most efficient at meeting that requirement, and then finally determining how often we need to test QC samples to be efficient and effective.

Let's start by discussing the quality requirement.

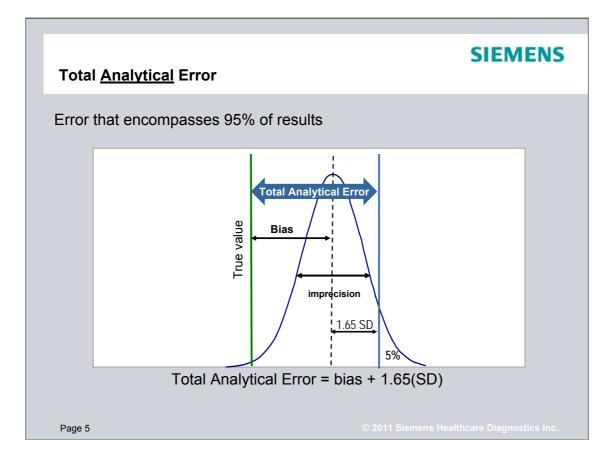


A key concept to think about when looking at QC results is what constitutes a real performance problem ? How do we know when there is a real meaningful problem with the method ?

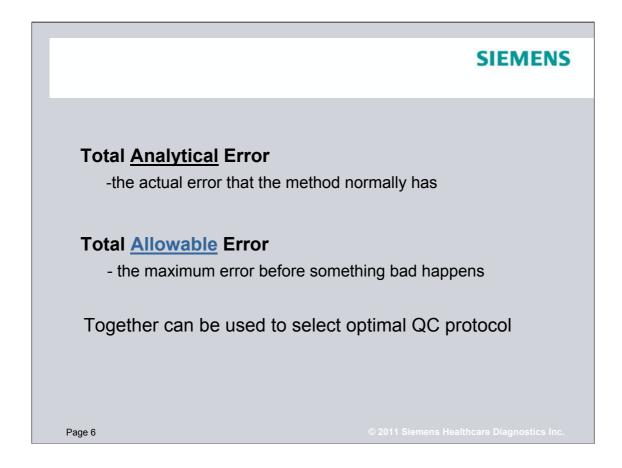
Is it always meaningful when the QC results fail a statistical QC rule? Or is the true criterion that method performance has changed enough to impact medical care. I think we all agree that the later is our real concern. However, historically we have mostly all assumed that these two things were equal and identical; that failure of a statistical QC rule **always** meant that there was a medically significant change in method performance. But is that always the case? Actually, we know it is not. This is reflected in the fact that we often report results even though the QC results have failed a rule. We say that if only one level is out, it's OK to report, or similar evaluations. Here we have an excerpt from the original "Westgard Rules" paper in 1981 that indicates that it is acceptable and necessary to sometimes recognize that just because QC results have failed a statistical rule, even the "Westgard Rules", it may still be reasonable and necessary to release results while we are investigating the rule failure.

Statistical QC rules are tools we use. Tools that help us to know that some degree of change has occurred in the method. Then that change needs to be put in perspective relative to our quality requirement for method performance. We have been doing this intuitively and informally for years.

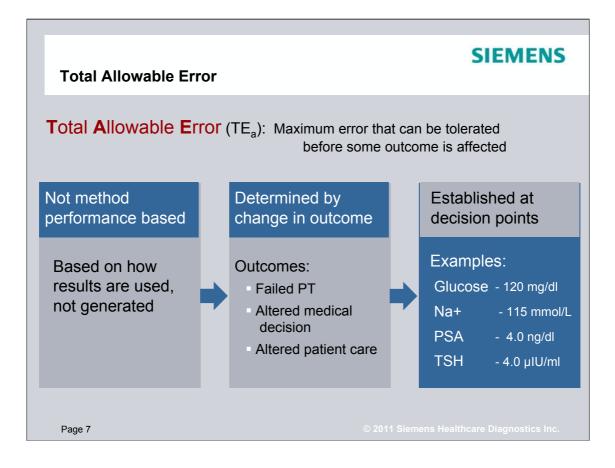
Let's look at formally establishing our quality requirement and see how QC rules really relate to it. To do this we want to introduce a two very useful concepts – Total Analytical Error and Total Allowable Error.



Total analytical error is defined as the error that encompasses 95% of the results for a given method. As we know, error is made up of two components. Constant error, often called bias, is the average consistent error seen over time. We would like to reduce or eliminate bias, but cannot always do so. The other component of total error is random error or imprecision. This is an inherent characteristic of the method. To estimate the total error for the method we want to capture the error that covers 95% of the results of the method. Since random variation is symmetrical around the mean, it sometimes adds to total error and sometimes reduces total error. Since we are only interested is the maximum error, we only look at the random error that increases total error. Using the SD as the measure of random error, the combined bias and imprecision that covers the error for 95% of results is bias plus 1.65 times the SD. That becomes our formula for estimating total analytical error.



So total analytical error is the actual error we have. We next want to look at what is the maximum error that can be tolerated before we impact patient care. There have been a number of ways to describe this, but the one that is currently most widely used is Total Allowable Error. Using these two concepts together ca help us design effective and efficient QC. Let's look at Total Allowable Error in more detail.



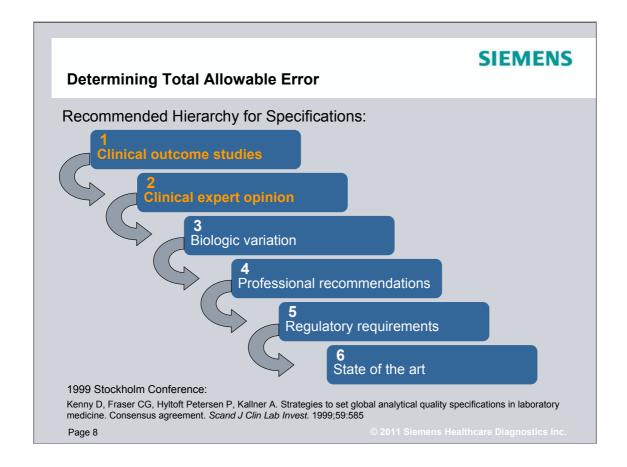
Total Allowable Error is the maximum error we can tolerate for an assay before some outcome like medical decision making or patient care is impacted.

Total allowable error is NOT based on current method performance. It is determined by how the results are used medically, not how the results are determined analytically. So it is independent of the method used. Since Total Allowable Error is dependent on the clinical use of the test result and the inherent biologic variability of the analyte, it is not the same for all analytes. Therefore it has to be established for each analyte and for each medically important concentration for the analyte. The total allowable error for calcium is the same regardless of what instrument or method is used to measure calcium.

The idea of total allowable error is that if we exceed it, the some outcome will be affected ... we may fail proficiency testing, a medical decision may be altered, patient care may be changed.

Since the concept of total allowable error revolves around medical decision making, typically we estimate the allowable error at concentrations where medical decision are made. To understand how this concept may be used let's try defining Total Allowable Error for a few specific analytes as examples. We'll use Glucose, Sodium, PSA and TSH. The first step for each analyte is to define a medically important concentration. For Glucose, 120 mg/dl is a decision point for the diagnosis of diabetes; for Sodium 115 mmol/L is the decision point for hyponatremia and severe electrolyte imbalance; for PSA a result above 4 ng/dl is suggestive of increased risk for cancer and should be followed up; and for TSH a result above 4.0 uIU/ml indicates possible hypothyroidism.

So, how do we decide what Total Allowable Error should be for a method? Many authorities discussed this for a number of years and in 1999 there was a conference held is Stockholm to develop a consensus approach.



At the conference it was recognized that there is not one simple approach that will work to define Total Allowable Error for all methods. So a hierarchy was developed to start with the most medically sound approaches and move to other approaches if the optimal was not possible. Here is that hierarchy.

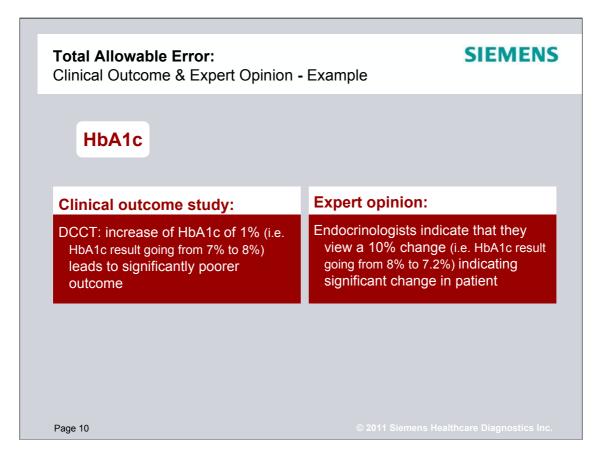
Let's start with looking at the use of outcome studies and Clinical Expert opinion

<b>Total Allowable Error:</b> Clinical Outcome & Expert Opinion	SIEMENS
How much change in a result alters medical outcome ? Becomes the total allowable error for that analyte Clinical outcome studies:	?
<ul> <li>Cardiac disease – Framingham, TIMI, Women's Health</li> <li>Diabetes – DCCT, NHANES,</li> <li>Large, prospective, long term studies looking at clinical</li> </ul>	
<ul> <li>Expert Opinion:</li> <li>Review institutional standardized care protocols</li> <li>Consult with physicians for expert opinion</li> </ul>	
Page 9 © 2011 Siemens	

We are trying to establish how much change in a result will alter medical decision making and patient care. That amount of change then becomes our allowable error since any change less than that will not cause a physician to make a different decision.

Clinical outcome studies are the optimal source for this information. They are focused on the decision making in specific medical scenarios, like diagnosis and management of heart disease or diabetes. These studies are prospective, long term studies that objectively assess how treatment decisions affect the outcome. Often lab results are used to make the treatment decisions. This is the most specific data we can use.

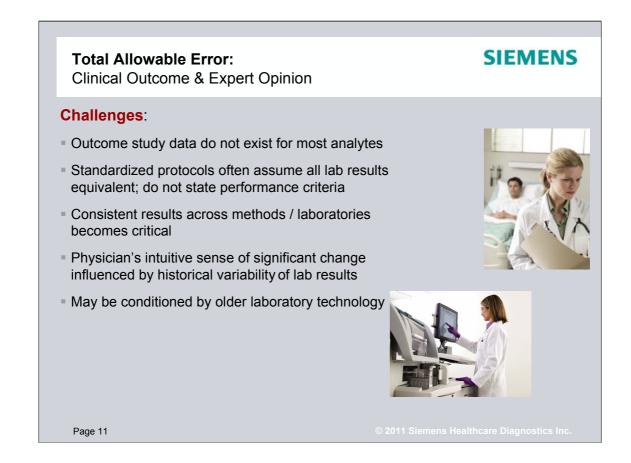
Using medical expert opinion may seem an obvious choice for setting Total Allowable Error. Essentially we want to know how much change in the result for a test will cause physicians to change their decision and that becomes the limit. To assess this we can look at the consensus derived standard treatment protocols that are used in many healthcare facilities today or we can consult with physicians. This will give us the benefit of their collective experience on how to best use lab results.



An example of a method with clinical outcomes-based data that can be used to make comparability recommendations is the use of the hemoglobin A1c assay (HbA1c) for monitoring an individual's diabetes control. The Diabetes Control and Complications Trial on Clinical Outcomes Related to HbA1c indicated that a HbA1c of 8.0% has a poorer clinical outcome compared to a HbA1c of 7.0%, and should therefore be accompanied by a change in patient management.

We can use the same example for Expert Clinical Opinion. A survey of endocrinologists might indicate that clinicians interpreted a 10% change (eg, a change in HbA1c concentration from 8.0% to 7.2%) in the HbA1c result as a significant change in a patient's clinical condition.[i]

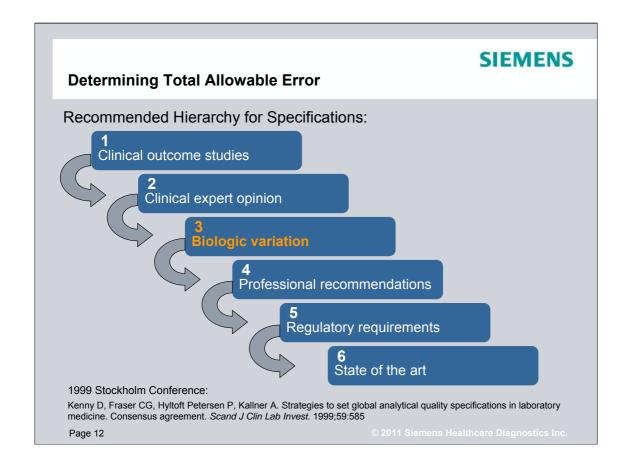
[i] Petersen PH, Larsen ML, Horder M. Prerequisites for the maintenance of a certain state of health by biochemical monitoring. In: Harris EK, Yasada T, eds. *Maintaining a Healthy State Within the Individual*. Amsterdam: Elsevier; 1987:147-158.



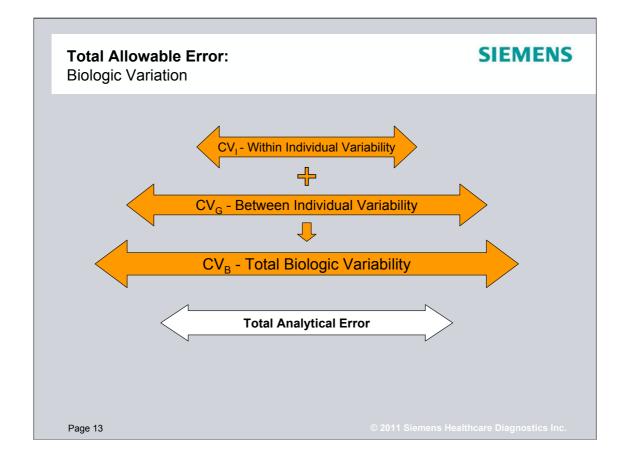
The challenge is that outcomes studies and clinical protocols don't exist for most analytes. So, while they may be useful guidance for some analytes, for most there is no standard of this type. Also, when soliciting expert opinion, how do you decide how much change is critical ?

It may seem straightforward to just consult with physicians about how much a Glucose result has to change before they would consider it significant, but there's a problem. Physician's intuitive sense of how much change is significant is in large part based on their experience with how variable lab results are compared to changes noted in the patient's status. This intuitive sense is shaped by the variability in lab results seen in the past and doesn't necessarily reflect current test performance.

So these approaches are very valuable, but may not be practical for all analytes.



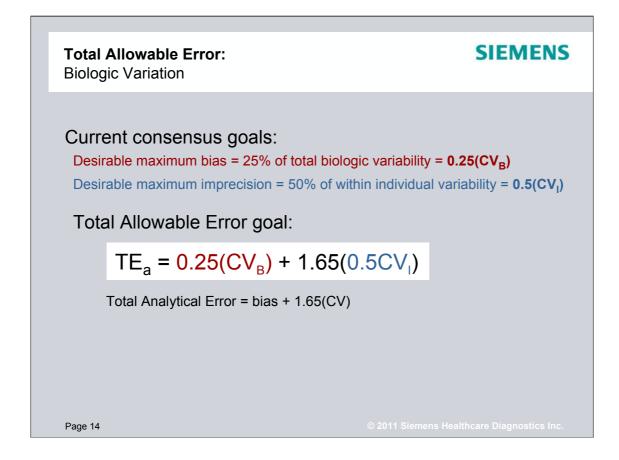
Now let's look at Biologic variation



All analytes we measure show inherent intra-individual variation. There is typically some degree daily variation that may follow a circadian rhythm. There are longer term variations including some seasonal. All of these variations are independent of any pathologic change and are part of normal physiology.

Further, the usual analyte concentration varies between individuals as well. As might be expected this is a larger variation than is seen within a single individual. The combination of intra-individual and between individual variation is called total biologic variation. We have all seen this variation reflected in the reference interval or "normal" range commonly used to interpret lab results. As most often used, the reference interval represents the central 95% of the range of values found in a population of healthy individuals.

The goal in using biologic variation to set total allowable error is that the analytical error should be small compared to the natural biologic variation. That way the analytical error will essentially be "lost" in the background noise. A number of articles have been published on how to achieve this and a consensus has emerged.

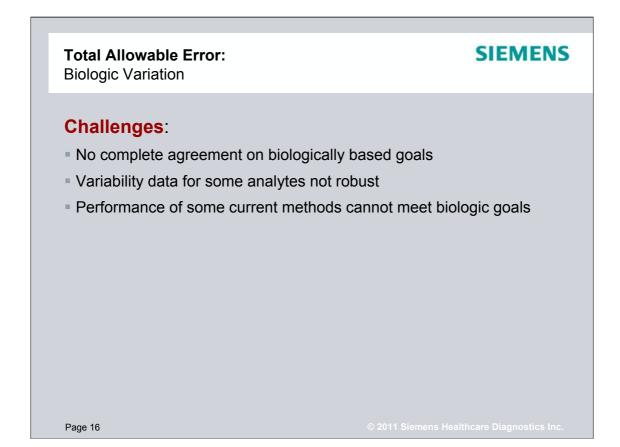


The current consensus on using biologic data to set analytical performance goals sets the limits of analytical error based on the biologic data. The desirable goal for bias is no more than 25% of total biologic variability. The desirable goal for imprecision is no more than 50% of within individual variability. Using these proposed limits, we can set a goal for Total Allowable Error that will encompass 95% of results for a given analyte.. The estimated Total Allowable Error is the bias plus 1.65 times imprecision. This is the current working model for estimating total allowable error from biologic data.

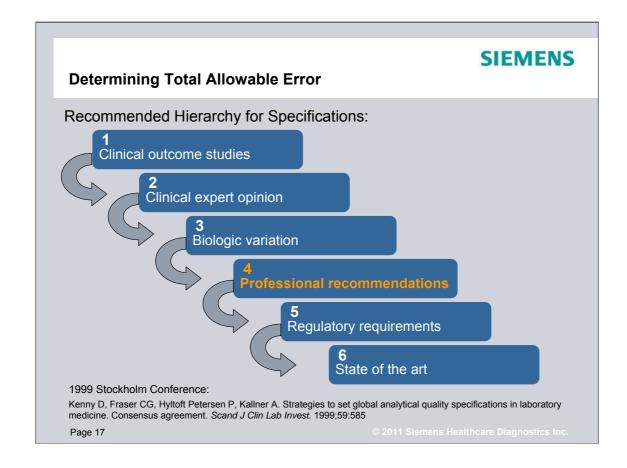
Desirable S	pecifications	for Total Err	or, Imprecisi	on, and Bia	as, Derived fr	om Biologic '	Variation
	rez V, Cava F, Ga biologic variation						"Current
,			bject CV values	of analytes a	and Desirable An	alytical Quality S	Specifications for
imprecision, b	ias and total erro	r					
		Biologic Variation Desirable Specifi		cation			
	Analyte	CV <sub>1</sub> (%)	CV <sub>B</sub> (%)	CV (%)	Bias (%)	TE <sub>a</sub> (%)	
	Glucose	5.7	6.9	2.9	2.2	6.9	
	Na <sup>+</sup>	0.7	1.0	0.4	0.3	0.9	
	PSA	18.1	72.4	9.1	18.7	33.6	
				01 500 100 1000 100 100 100 1000 100 100		53 1000 1000 1000 1000 1000 1000 1000 10	

To look at some of these biologically based goals, an excellent resource is an ongoing series of articles published by Carmen Ricos and colleagues. The table of biologic data can be readily accessed at Dr. Westgard's website.

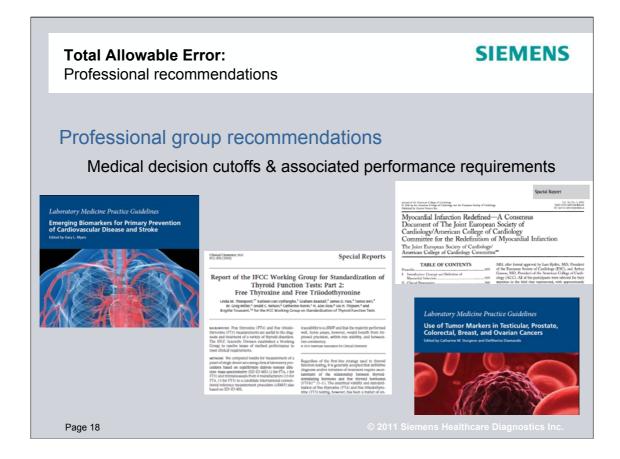
Here we can see the data for our four example assays with the Total Allowable Error goal listed in the right most column.



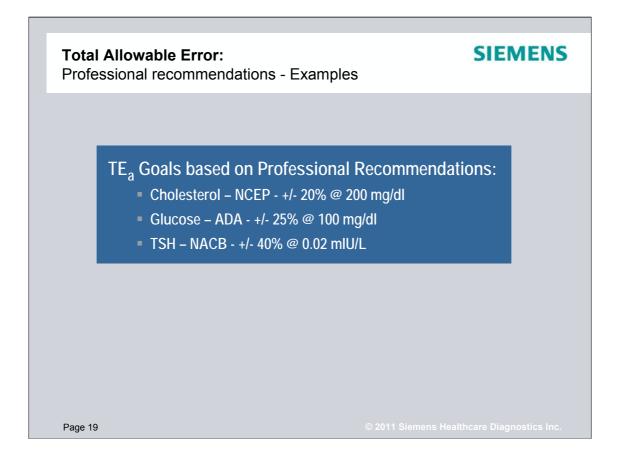
There are challenges. First this is a consensus model, which implies some degree of disagreement on how the goals should be set. Second, the data used to determine biologic variability is not robust for all analytes. We have excellent data for many analytes, but the data is not as solid for many others. Finally, some methods in current use cannot achieve the level of performance necessary to meet goals set using this model. Current technology is not capable. Example analytes where this is an issue are Sodium and often Calcium.



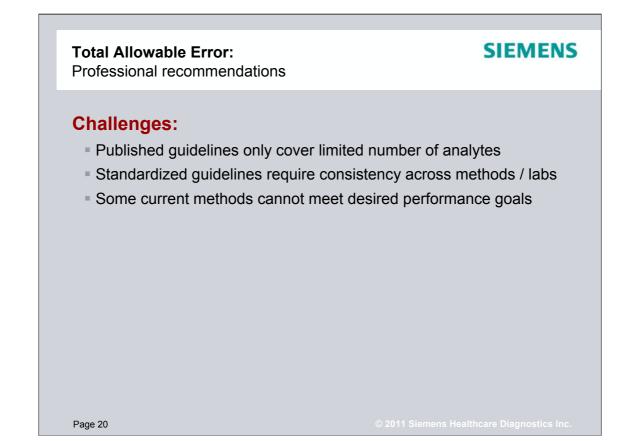
Another option is to use recommendation made by professional groups



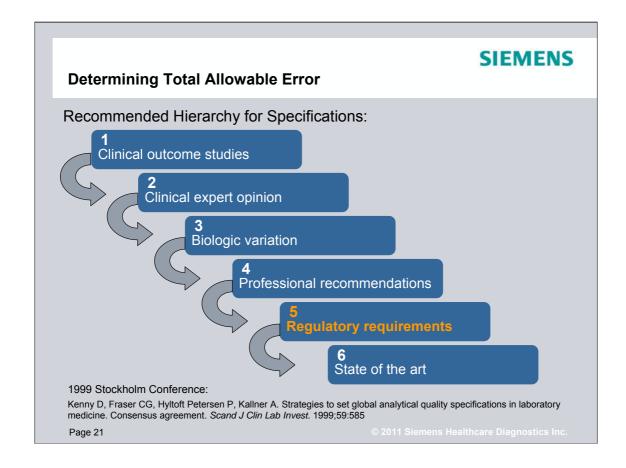
There have also been a number of published studies and reports by professional groups that also establish specific medical decision points for some analytes. In these studies and reports, tolerable error limits are often also defined. These reports can be very useful in establishing Total Allowable Error for those analytes. Since the data used to establish the recommended performance criteria are not always outcome based, the recommendations in these reports are not a solid as those from outcome studies. These reports are based on outcome data whenever possible, but, as we already indicated, that data does not exist for many analytes.



Examples include the National Cholesterol Education Program (NCEP) establishing that cholesterol results changing by more than 20% at 200 mg/dl is clinically important, or the American Academy of Cardiology (ACC) indicating that the decision point for Troponin should be the 99<sup>th</sup> percentile of the healthy population and that the maximum allowable error at that concentration is 20%, or the National Academy of Clinical Biochemistry (NACB) publishing guidelines for thyroid testing that indicate that a change in TSH of more than 40% at 0.02 mIU/L is significant.



As with the other approaches discussed, one limitation is that these reports and recommendations do not exist for all analytes. Virtually none of these protocols, studies or reports make any allowances for lab to lab or method to method differences in results. None suggest interpretation of results using lab specific reference intervals. This means that there is increasing pressure on manufacturers and laboratories to minimize or eliminate these differences. As we all know this is not simple task for a number of analytes, but progress is being made and will continue to be made. Finally, these recommendations are clinically based and focus on what is desirable clinically. There have been a couple of cases where the performance recommendation cannot be met by any method in current use. Technology has not caught up with the perceived medical need.



Next in the hierarchy are the recommendation made by regulatory agencies.



Agencies in many countries and even state agencies here in the US manage External Quality Assessment (EQA) or Proficiency Testing (PT) programs and have established acceptable performance limits for these inter-laboratory testing programs. If these limits are used to establish Total Allowable Error, we can then set as a goal detecting any change in method performance that would cause a failure with an EQA or PT result.

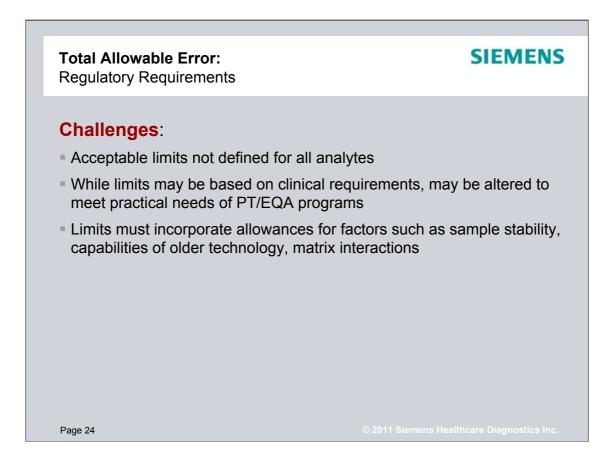
This approach to establishing Total Allowable Error has been very popular in most of the literature articles about Total Allowable Error and these limits are often listed in tables in these articles and in some software as recommended values for TEa. Let's look at some examples.

Total Allowable El Regulatory Require		nples	SIEMENS
Less than half the	mples in literat typical laborate ommittee cons	ure and software for ory menu of analytes sensus based on 198	Total Allowable Error limits has CLIA PT goals
	<u>Total Allowa</u> Glucose Sodium PSA TSH	i <u>ble Error based or</u> 120 mg/dl ± 10% <u>115 mmo</u> l/L ± 3. <u>None</u> 4.0 μIU/ml ±21	6 47%
Page 23		© 2011 Si	emens Healthcare Diagnostics Inc.

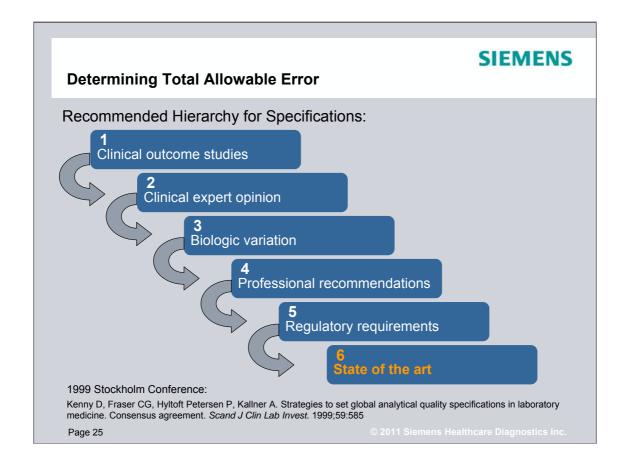
In the US the CLIA regulations have established performance criteria for a number of analytes.

Here are the CLIA goals for our example analytes With our example, we can find CLIA goals for Glucose, Sodium, and TSH and we can use the goals to set Total Allowable Error specifications at our chosen medical decision points. However there is no CLIA performance goal for PSA as is the case for many analytes and most immunoassays.

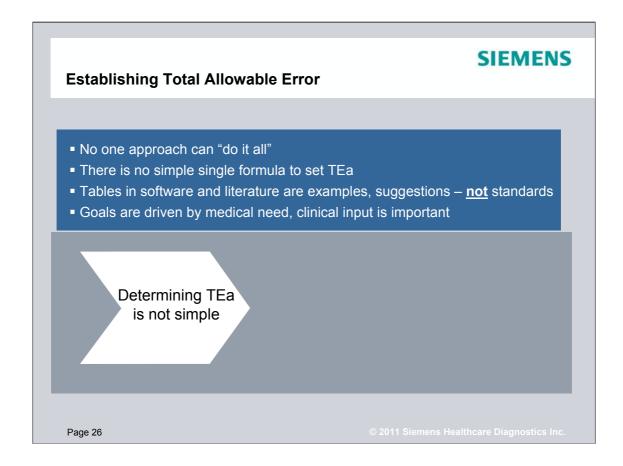
These performance goals can be used as the Total Allowable Error goal. However, these CLIA performance goals were established prior to 1992 using a consensus process and are based on the expected performance of analytical systems in use at that time. They don't reflect well current performance or necessarily medical needs and, most importantly, the goals are only set for about 40 analytes. These goals can be a good resource when establishing Total Allowable Error, but they are not a gold standard.



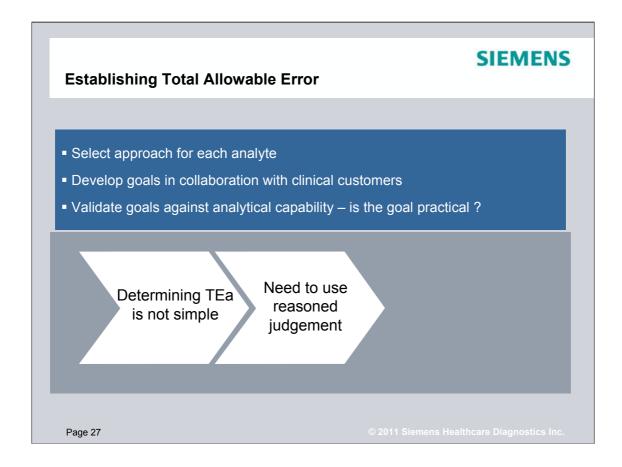
There are some challenges to using EQA or PT limits for Total Allowable Error. Especially in the US many analytes commonly part of the labs menu do not have CLIA defined limits. Further, while these limits are often based on medical usefulness criteria, the actual limits are modified to meet the needs of the EQA / PT program. Things like sample stability, possible matrix interactions, the need to cover a wide range of analytical technology, etc. often drive adjustment of the medically derived limits to meet the practical constraints of an EQA / PT program.



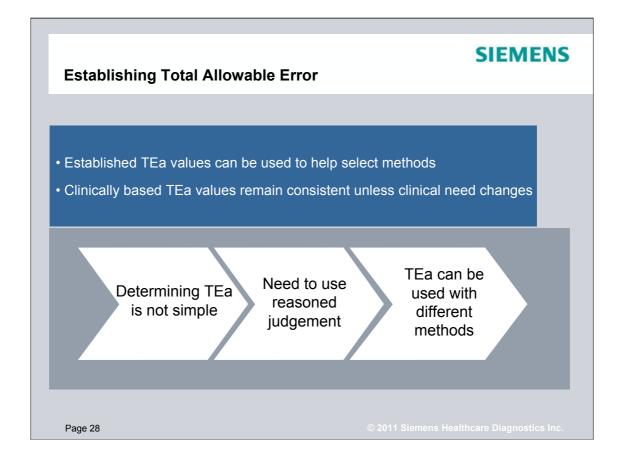
Finally we have performance goals based on the current performance of the available methods. This is the final default criteria if nothing better can be found. This should be our last resort, not our first choice.



Setting Total Allowable Error goals for all analytes is the hardest part of developing an efficient and effective QC protocol. There is no simple, one right way to estimate Total Allowable Error. The example tables from literature articles are just that, examples. They are not standards or necessarily the best approach for us to use. We need to keep in mind that the goals are primarily clinical in nature.

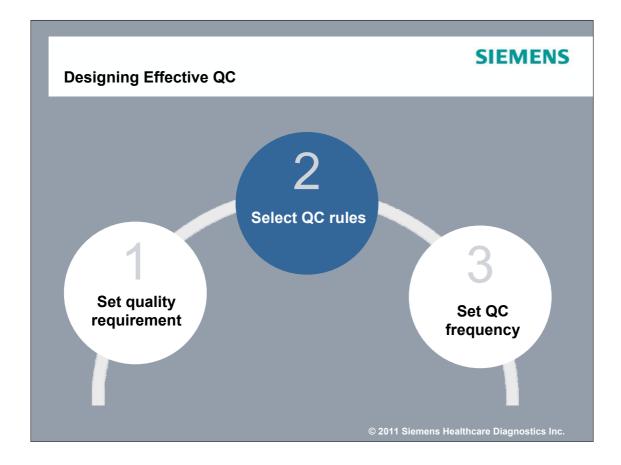


We need to use a combination of approaches and work in collaboration with our clinical colleagues to establish our allowable error goals. Then, once we have a proposed set of goals, we need to validate them against the capability of our instruments and methods. It does no good to set a performance goal that no instrument or method can achieve. We always have to balance what we would like against what is possible.



Establishing total allowable error goals for all analytes in the lab takes a fair amount of time and effort and is never easy. However, once it is done. It is essentially done for all time. Since the Total Allowable Error goals are not based on how current methods perform, but rather on how results are used, once the goals are agreed on they can be used for a long time with different instrument systems. So, in the long run, the effort to set these goals is worth it.

Once we have established our Total allowable Error, we can use it to help select the optimal QC rules for our analytes ...



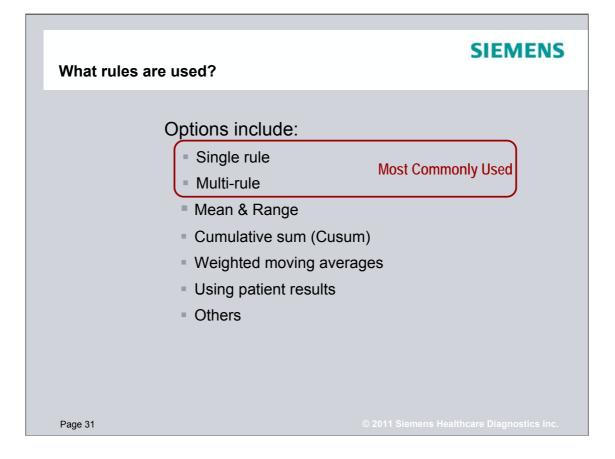
Once we have established our quality requirement, we can use this to select our QC rules. However, before we discuss how to use the quality requirement in rule selection, we need to review some key concepts behind the function of QC rules so the selection process makes sense.

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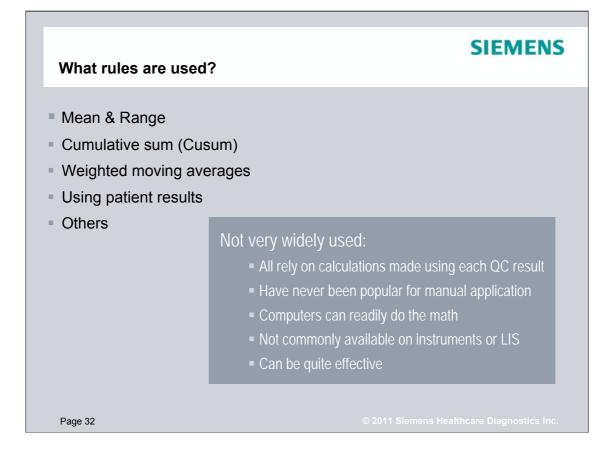
An initial step in selecting rules is to decide which type of rules we want to use. There are two basic types. Failure rules are designed so that, if the QC data fails the rule, we say the method is out of control and we stop reporting results until further action is taken. Clearly this is the most common type of QC rule and all QC protocols need to be based on one or more failure rules.

We also have warning rules. These are rules that typically have too high a false positive rate to be effective failure rules, but they can function very well to give us an early indication that some change in performance may be occurring and allow time to investigate before we trip the failure rule. With a warning rule, if the QC results fail the rule, WE do NOT stop releasing results. Instead, we recognize that the method is still acceptable, but something may be happening. So we start to investigate to see if there really is an issue without interrupting the work flow.

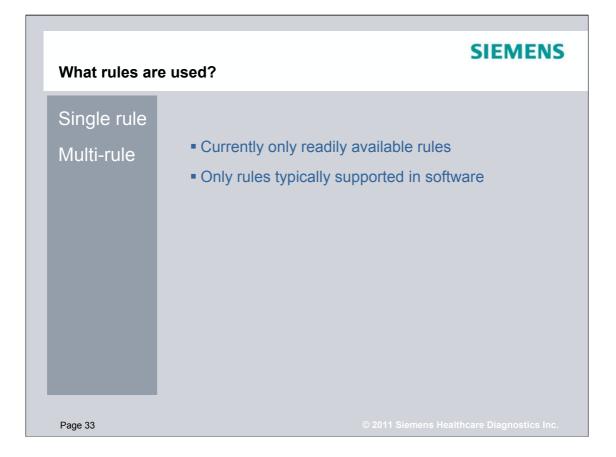
The usual problem with using warning rules is that pretty soon everyone starts treating them as failure rules and stops reporting when the warning rule trips. This negates the whole idea of a warning rule and makes the QC process very inefficient.



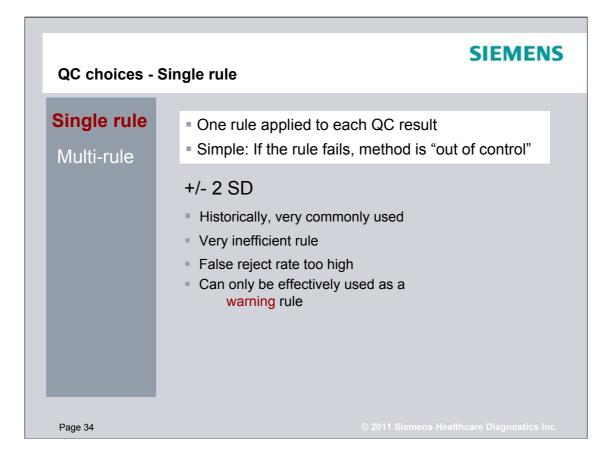
So what rules are available to us to use. There are actually quite a lot of options. Single rule protocols and multi-rule protocols are the most commonly used and we will discuss them in some detail.



These other options: Mean & Range, cumulative sum, weighted moving averages, using patient results and others like multi-variate approaches are all well documented in the literature and can be very effective and efficient. They have not been widely used because they pretty much all require that calculations be made each time a QC result is evaluated. In the past this was not practical for many labs. However, now all the instruments have powerful computers, many labs use middleware products that use powerful computers and most all labs are connected to LIS systems that can perform the calculations. However, if we look at the QC support software on our instruments, our middleware, and our LIS systems, we don't find these options available. So we are still left with single rule and multi-rule protocols as the most practical because they are well supported.

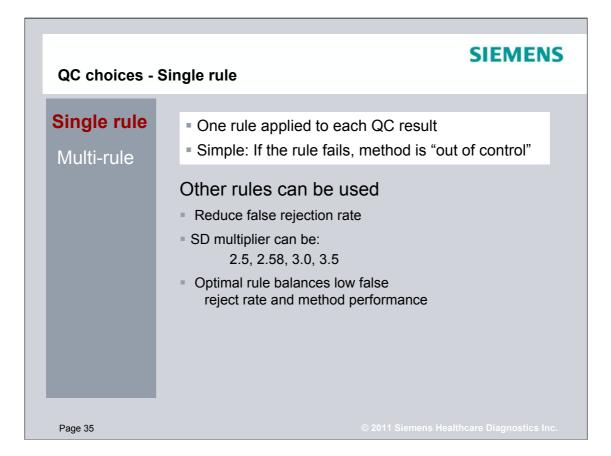


For now we will focus on Single rule protocols and multi-rule protocols since these are the most readily available procotols and the only ones generally supported by the software we use to manage QC.



A single rule protocol is just what it sounds like. A single QC rule is applies to each QC result as it is generated. If the result fails the rule, the method is deemed "out of control".

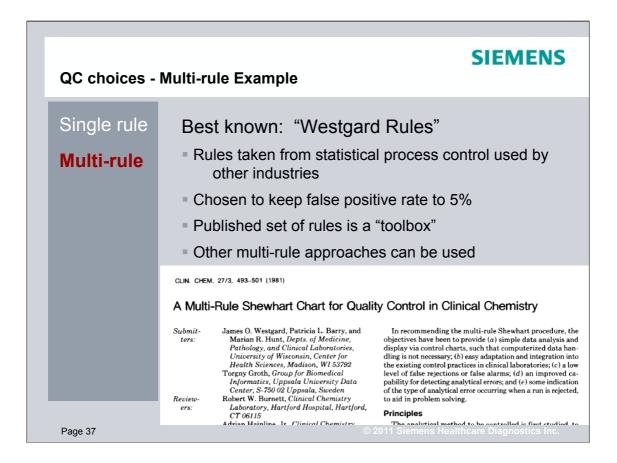
Historically, the single rule +/- 2 SD has been the most commonly used. This is a very inefficient rule due to it's very high false positive failure rate, especially when used with multi-level QC material for for many methods concurrently. It can be an effective warning rule ... but most of the time the warning rule is actually used as a failure rule and we have gained nothing.



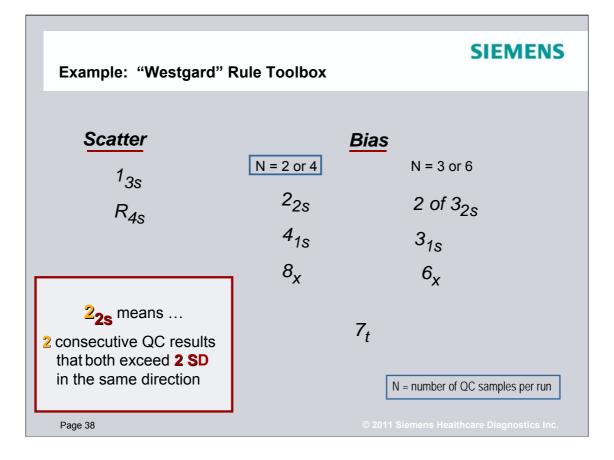
However, +/- 2 SD is not the only possible single rule. SD multipliers like 2.5, 2.58, 3 and even 3.5 can be effectively used to control the false positive rate and reliably detect change in performance. Notice that it is not required by statistics or science that the multiplier of the SD used for a single rule needs to be a whole number. The only reason most rules used historically have been whole numbers is that those rules were developed when we were doing the math in our heads.... And whole numbers are easier to work with. Today with computers doing the math, the multiplier can be any value we want in order to get the detection or false positive rate we desire. A multiplier of 2.58 SD gives us a false positive rate of exactly 1% per method per control. What is critical is to match the choice of rule to the quality requirement and the usual performance of the method. We will look at this in detail in a bit.

QC choices -	Multi-rule QC protocols
Single rule Multi-rule	<ul> <li>Series of rules to validate QC results</li> <li>If one or more rules fail – method is "out of control"</li> <li>Each rule alone may not be ideal; taken together they provide effective QC monitoring</li> <li>Rules designed to assist in detecting trends</li> <li>Details are important: rules must be used exactly as designed</li> </ul>
Page 36	

The other commonly available choice for QC rules is a multi-rule protocol. As the name implies, multi-rule protocols use a series of several rules to evaluate QC results. If the QC result fails one or more of the rules, the method is deemed "out of control". The individual rules used are selected to have very low false positive rates. As a consequence, they often focus on specific types of errors and, each used alone may not be completely effective in detecting all changes in performance . However, used together, they reliably detect changes with a low overall false positive rate and ... looking at which of the rules failed can often provide useful information on possible root cause. Key to using these rules is understanding that each rule must be applied in a very specific way to be effective. We have to pay attention to these specific requirements or we may lose the overall effectiveness of the process. Details are important.

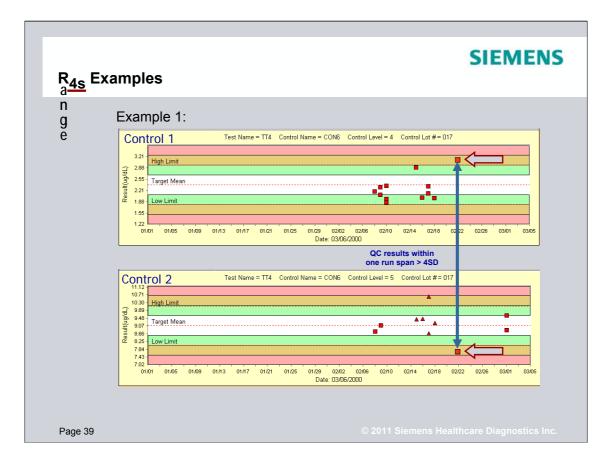


The best example of a multi-rule protocol is also the most widely known, the so called "Westgard Rules". Dr. Westgard and three other authors published the paper introducing these rules 30 years ago. The rules used were selected from statistical control rules used in other industries. Dr. Westgard selected rules that would best fit the way a clinical laboratory operates and which would have a very low false positive rate. The rules as described in the riginal article and in all subsequent writings are not a fixed set of required rules, but rather a tool box of rules that can be used. There are other multi-rule protocols available, but they all essentially work the same way. Let's look at Dr. Westgard's proposed rules in a little more detail.

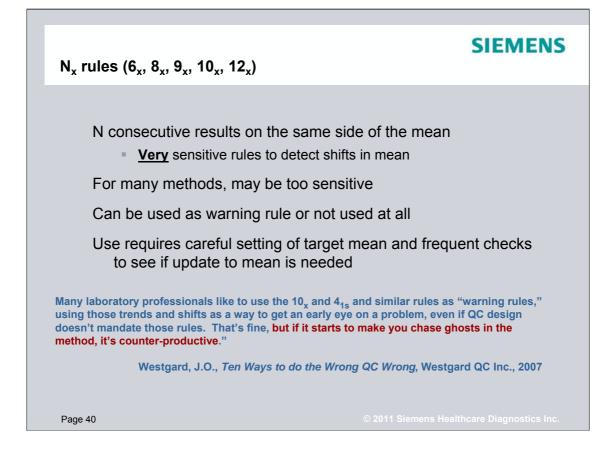


Here is Dr. Westgard's rule tool box. As you can see, some rules are designed to detect increased scatter or imprecision. Other rules are designed to detects changes in bias or shifts. We can also see that which rules you should use depends in part on how many QC results are being evaluated together. We have one set of rules for when we use 2 levels of controls and a somewhat different set if we use three levels of controls.

The notation may seem strage at first, but it is easy to understand. 2 2s means two consecutive QC results that both exceed 2 SD on the same side of the mean. Similarly 4 1s would mean 4 consecutive QC results all exceeding 1 SD on the same side of the mean. Let's look at these other rules

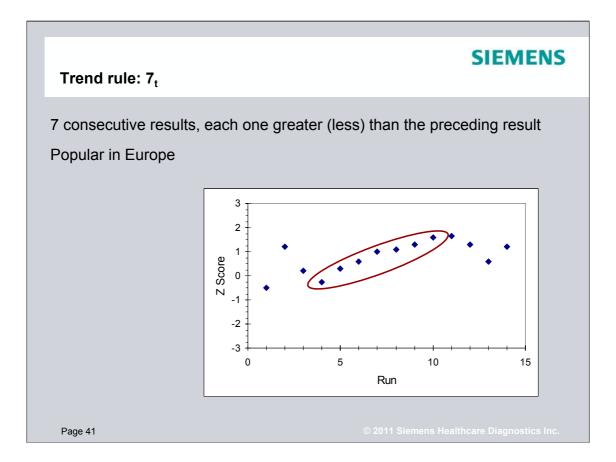


The R 4 s rule looks at the range spanned by two controls within the same run. If the span exceeds 4 SD, then the rule fails. Note this applies only to controls run together in a single run and that they do not have to be consecutive. If we are using three levels of control, if any two of the three results show a span exceeding 4 SD, then the rule fails.

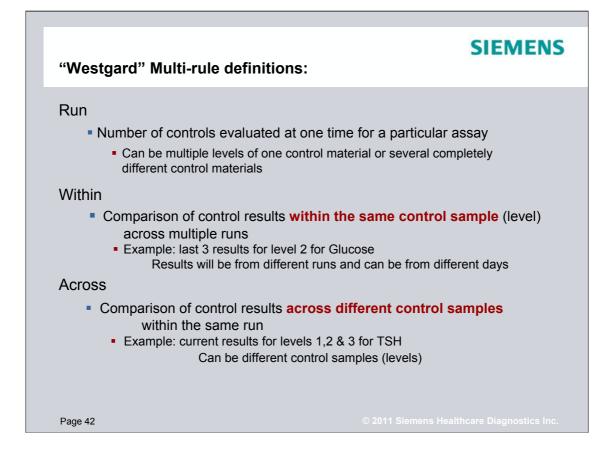


The "N x" rules are interesting. The basis of the rule is "N" consecutive QC results all on the same side of the mean. Values used for "N" have been 6, 8, 9, 10, 12. These rules are designed to detect changes in bias or shifts and they are very sensitive ... sometime too sensitive. Dr. Westgard has recommended using this type of rule as a warning rule in most cases or not using it at all. These rules are best saved for methods where there is little room for change in method performance. That is a very small minority of methods as we shall see.

If use of these rules is contemplated, it is absolutely critical that the target mean be carefully set using data from the instrument and that the mean be checked and updated regularly. One of the fastest routes to frustration and highly inefficient QC is to try to use these rules with QC targets taken from a package insert or IFU. That will virtually never work because the actual instrument mean almost never matches the IFU mean exactly. This is normal and expected, but if the IFU mean is used as a QC target mean, these "Nx" rules will consistently fail because of that difference.

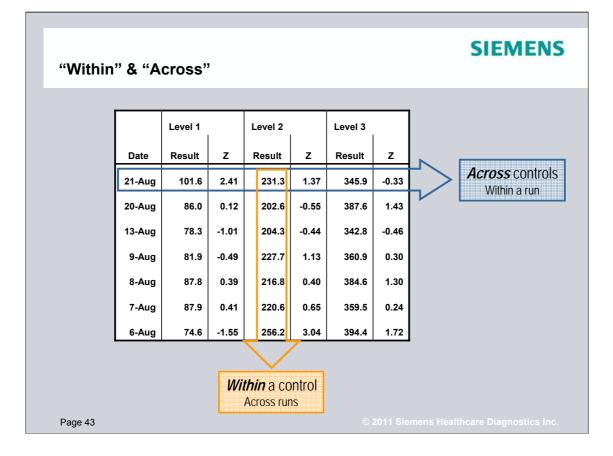


The last rule we will look at is the 7T. This is a trend rule that has been popular in Europe. It requires that 7 consecutive QC results each be greater than (or less than) the result immediately before. This is not the same as the "N x" rules since they only require that the results be on the same side of the mean. Here each result must be numerically greater than, or less than the one before it.

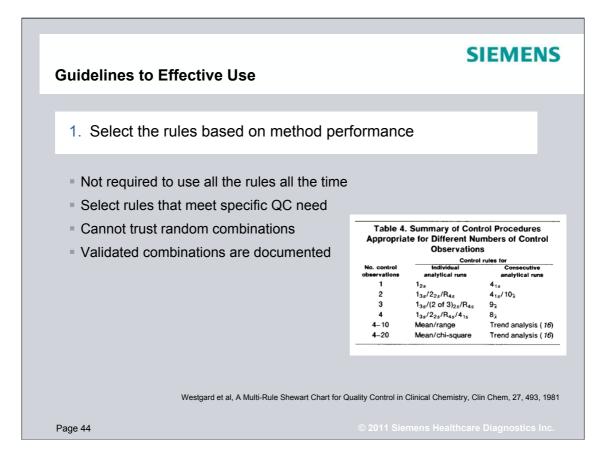


There are also a couple of concepts or definitions that are important to effectively using the Westgard Rules. The first concept is the "run". The term comes from the days when patient samples used to be tested in separate defined batches or runs, rather than continuously. As applied to these QC rules, the concept of run really iss about how many QC results will be evaluated together at one time. If we use a bi-level QC material and run both levels together, then the run is 2 QC samples and the rules are applied to both results simultaneously once both results are available.

The other two concepts are "within" and "across". These terms indicate how the rules are applied to the QC results. As originally used by Dr. Westgard, "within" refers to applying the rules within a single control material, like BioRad level 1. This often means looking back to previous QC runs on other days to have enough data to apply the rule. "Across" implies applying the ruule across different control materials within a single QC run. This would be applying the rule to the two QC levels run just now.

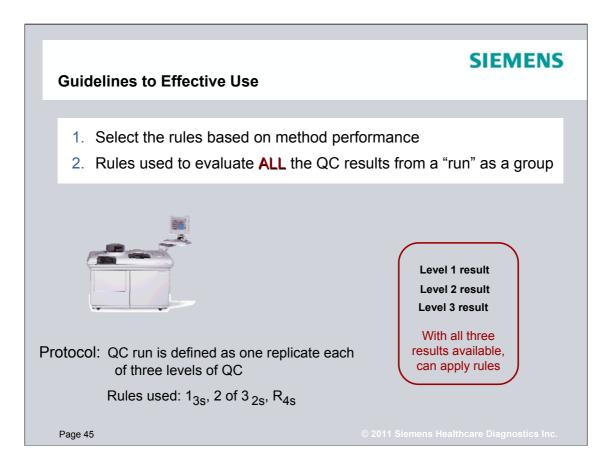


Here is another way to visualize the concepts of within and across. Most QC rules are designed to be applied both ways The idea behind looking back to previous days is to gain sensitivity to detect changes early on by using more data. This is really what we instinctively do when we look at the QC graph and review the data from previous days. Applying the rules this way just makes that look back part of the QC rules.



Now let's look at some guidelines to the effective use of the Westgard Rules

First – Select the rules used based on method performance. We will discuss how to do this is detail in a few moments, but right now I want to make the point that ... you are not required to use all the rules all the time. Even in the original paper that so many have referred to, Dr. Westgard selected the which subset of the rules to use based on the number of QC samples tested in each run. Today, the selection is driven by method performance. Key is that random combinations do not work. The rules have been validated to work in some very specific groupings. The specific groupings can be readily found on Dr. Westgard's website and even in the original paper as shown here.



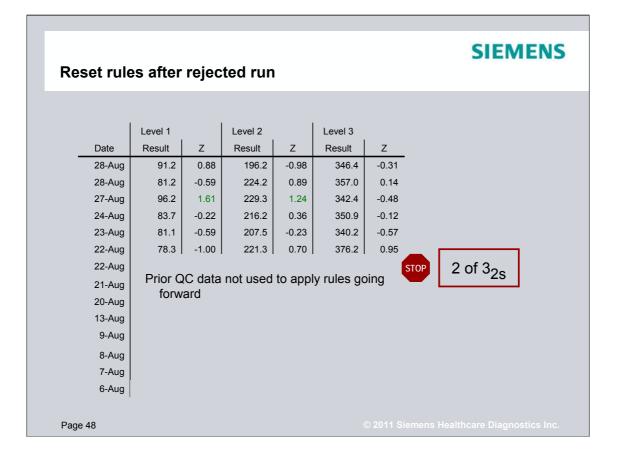
Next, the rules are designed to be applied to QC results as a Run ... not to each individual QC result as it is generated. This clearly becomes critical for a rule like 2 of 3 2S. If you don't have all three QC results, how can you apply the rule. This has rarely been an issue when people were manually applying the rules, but it can be an issue with computerized applications.

Guidelines to Effective Use	SIEMENS
1. Select the rules based on method p	erformance
2. Rules used to evaluate ALL the QC	results from a "run" as a group
<ol> <li>Once a "run" fails, future "runs" are results obtained after the rejec</li> </ol>	
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Once we have a rule failure, the data used to evaluate the rules cannot come from prior to the rule failure. Let's see how this works

### SIEMENS Reset rules after rejected run Level 1 Level 2 Level 3 Date Result Ζ Result Ζ Result Ζ 28-Aug 91.2 0.88 346.4 -0.31 196.2 -0.98 81.2 -0.59 224.2 0.89 357.0 0.14 28-Aug 27-Aug 1.61 342.4 96.2 229.3 1.24 -0.48 24-Aug 83.7 -0.22 216.2 0.36 350.9 -0.12 23-Aug 81.1 -0.59 207.5 -0.23 340.2 -0.57 22-Aug 78.3 -1.00 221.3 0.70 376.2 0.95 2 of 3<sub>2s</sub> 99.8 2.13 212.4 0.10 357.9 0.18 STOP 22-Aug 21-Aug 101.6 2.41 231.3 1.37 345.9 -0.33 20-Aug 86.0 0.12 202.6 -0.55 387.6 1.43 13-Aug 78.3 -1.01 204.3 -0.44 342.8 -0.46 9-Aug 81.9 -0.49 227.7 1.13 360.9 0.30 0.39 216.8 0.40 384.6 8-Aug 87.8 1.30 87.9 0.41 220.6 0.65 359.5 0.24 7-Aug 6-Aug Page 47

Once we have a failed run, we start over with the data used for rules going forward. So it will be 4 runs into the future before we can apply the 4 1s rule within a single control. However, this only applies to the QC rules. When we use this data to calculate a mean or SD, we use all the data except from the specific run that had the problem.



Once we have a failed run, we start over with the data used for rules going forward. So it will be 4 runs into the future before we can apply the 4 1s rule within a single control. However, this only applies to the QC rules. When we use this data to calculate a mean or SD, we use all the data except from the specific run that had the problem.

Guidelines to Effective Use	
1. Select the rules based on method p	erformance
2. Rules used to evaluate ALL the QC	results from a "run" as a group
3. Once a "run" fails, future "runs" are o obtained after the failed "run"	evaluated using only results
4. Rules were selected for manual app	plication to QC run in a batch

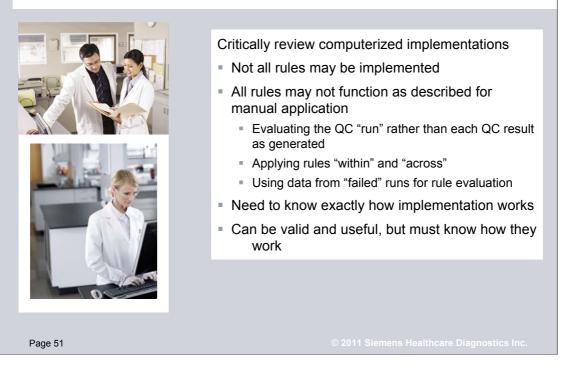
Recognize that the rules were developed in the 1970's and were designed to be realtively simple for people to use manually

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It is not difficult to teach someone to manually look at graphed QC results and apply the Westgard rules. Keep in mind they were always meant to be evaluated looking at a QC graph. It was never intended that anyone would try to use the rules looking at columns of numbers on a page.

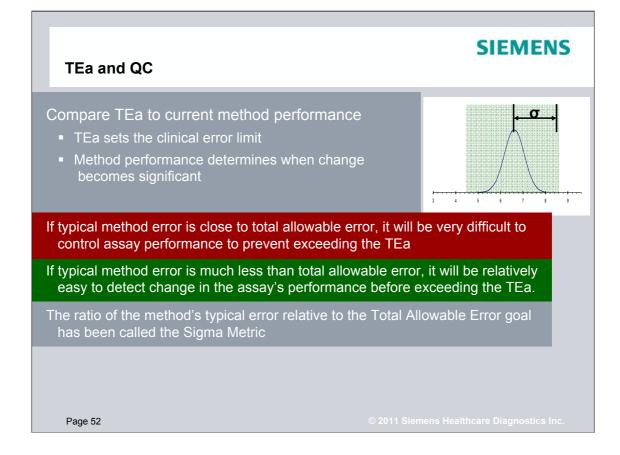
# SIEMENS

### Manual vs. Computerized rules



However, now most folks use some sort of a computerized implementation of the rules and there's the challenge. Most computer implementation of the Westgard Rules do not use the rules the way Dr. Westgard originally intended. Frequently not all the rules are available, especially those for three levels of control. Then the rules are often not applied "within" and "across" and finally the rules are often applied to each individual QC result as it is generated rather than collectively to the run.

These differences do not mean that these implementations of the rules are not good and do not work. They can be effective and do the job, but it is important that we know exactly how they work and not assume that just because they are called Westgard Rules, they are exactly as described in the original paper.

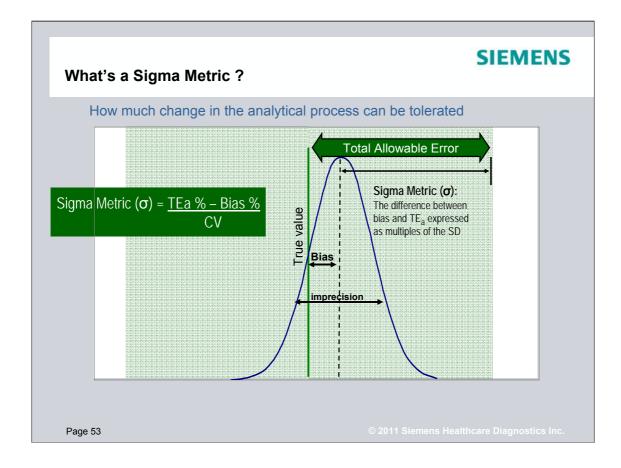


Now, finally let's bring it all together and use our Total Allowable Error based quality requirement and our understanding of the QC rules to see how we can select effective and efficient QC rules for our methods

To do this we compare our TEa goals to the actual performance of our methods on the instrument we are using. This is where we make the connection between TEa goals and actual method performance.

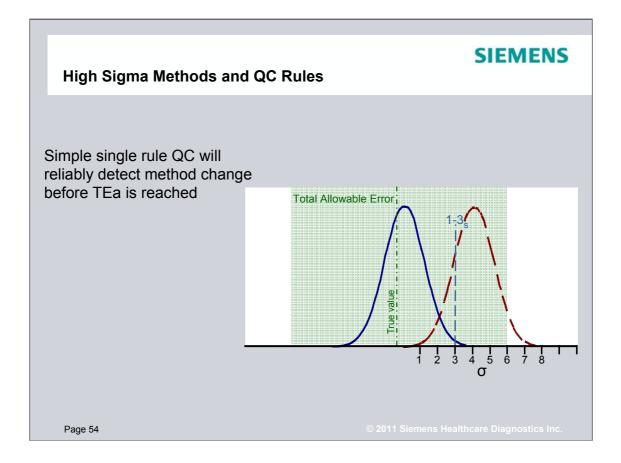
So, if Total Allowable Error is close to the actual performance of the assay, it may be difficult to monitor the assay and control it to prevent change in assay performance from impacting assay interpretation. However, if the actual method variability is small compared to the performance goal it will be easy to detect change in performance before it has an impact on patient care.

Recently, folks have begun taking the ratio of TEa to the method's variability as a guide to selecting QC rules. This ratio is called the Sigma Metric.

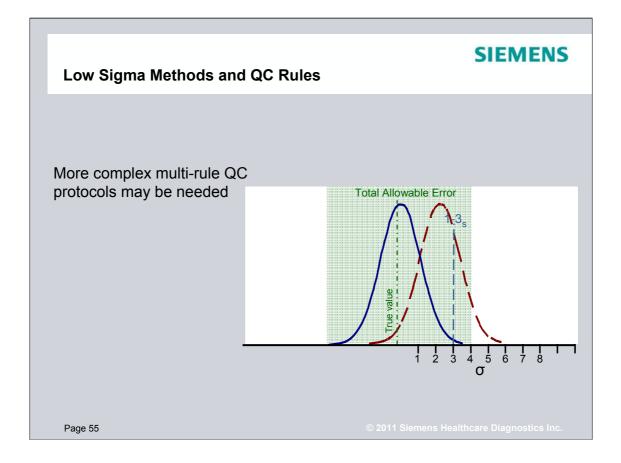


The Sigma metric is a measure of the difference between the actual method error and the Total Allowable Error. Here we see the performance of an assay relative to the "true" value and the Total Allowable Error. The Sigma Metric is calculated by subtracting the assay's bias from the Total Allowable Error goal and then dividing that difference by the assay CV. This gives the difference between current assay performance and the error goal as multiples of the CV or SD.

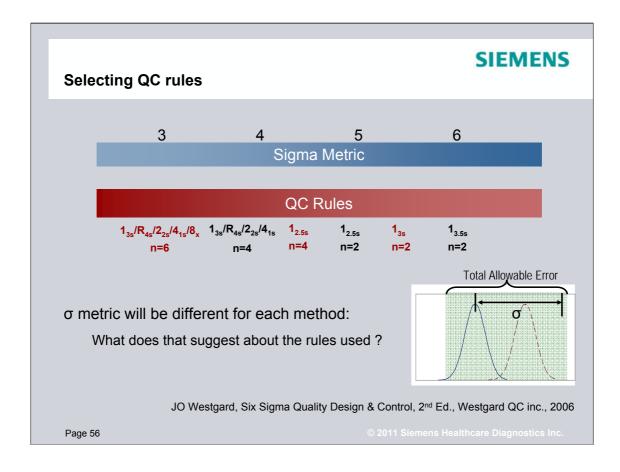
As you might expect, the ideal is for the Sigma Metric to be 6 or higher. Let's see how we can use this value to determine what QC rules will be effective.



With high sigma methods, the difference between typical performance and the total allowable error limit is sufficiently large that a simple single rule protocol like +/- 3 SD can readily catch any significant change in method performance before we exceed the allowable limit and still have a very low false positive rate.



On the other hand, a low sigma method doesn't have the same cushion to work with. In this case using +/- 3 SD will not be effective because we will have exceeded the error limit well before a 3 SD limit will consistently indicate the change in performance. In this case a multi-rule protocol will be more effective and which rules to use will depend on the sigma.



When we use the sigma metric to help select QC rules we find there is a continuum of which QC rules work best at which sigma metric.

If the assay's sigma metric is 5 or greater, it becomes fairly easy to detect change in performance before the analytical performance can impact decision making and the QC protocol used can be very simple.

If the assay's sigma metric is between 4 and 5- it's still fairly easy to catch change, but slightly more powerful QC rules are needed

If the assay's sigma metric is between 3 and 4 – it is more difficult to catch performance changes before they impact decision making, but it is still practical with reasonable QC protocols. The closer we are to 3 sigma the more complex the rule set.

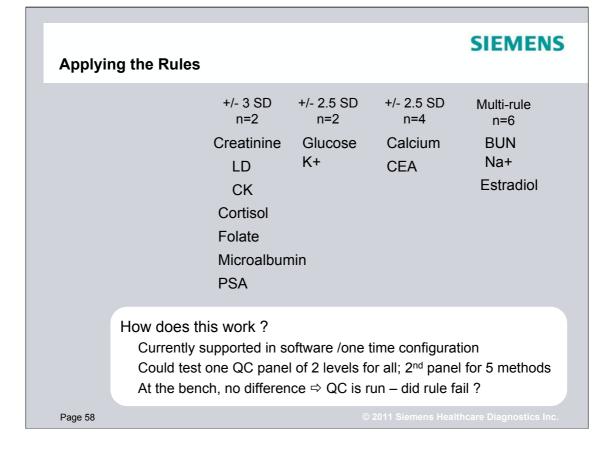
If the Sigma metric is less than 3, we need all the QC rule support we can get and even that may not be able to effectively monitor changes in the assay's performance to prevent any impact on decision making using statistical QC protocols alone.

Fortunately, most current assays fall into the 4 or better sigma range and so are OK. However, in the menu of almost every system are a few that do not. If that's the case, and an alternate better method is not practical, then we have to use maximum statistical QC and know that even that may not detect all significant changes.

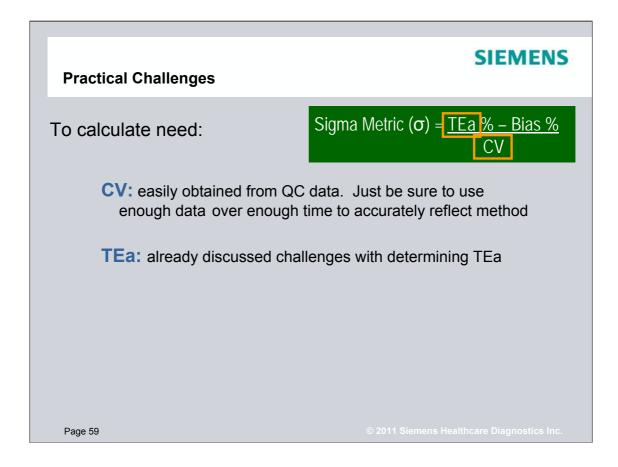
Since these choices are based on Sigma, it seems to suggest we could have multiple QC protocols in the lab

Applying the F	Rules				SIEMENS
Method Glucose Creatinine BUN K+ Na+ Calcium LD CK CEA Cortisol Estradiol Folate Microalbumin PSA	σ         4.8         7.5         3.3         5.0         2.9         4.5         6.2         9.5         4.0         6.2         3.4         6.9         9.2         6.1	+/- 3 SD n=2	+/- 2.5 SD n=2	+/- 2.5 SD n=4	Multi-rule n=6
Page 57			C	2011 Siemens Heal	thcare Diagnostics Inc.

When you do a sigma analysis and look at the results, it's easy to see that we will certainly not use the same QC protocol for everything in the lab and probably not even for all the methods on a single instrument. How are we supposed to manage that ?!



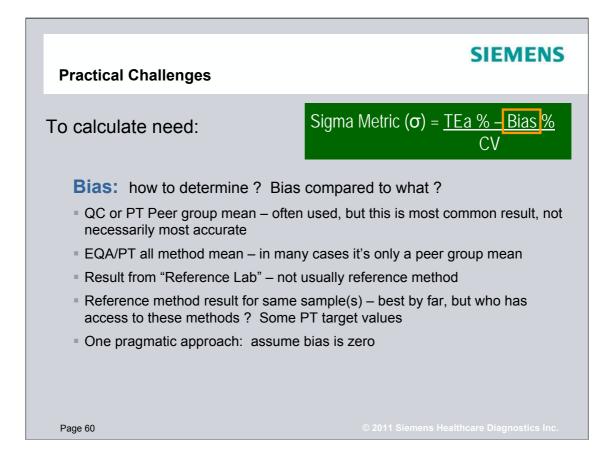
As we work it through we can see that methods get grouped into one of three or four different QC protocols based on their sigma value. So we only have a small number of different QC protocols. Still who can remember this? No one can or needs to. The QC software on most instruments today allows QC rules to be assigned on a method by method basis. A number of Siemens systems have supported this for more than 10 years. So we don't have to remember, the computer does. We configure the QC software one time and it remembers from that point on. Then we can use QC panels to easily schedule the number of QC samples appropriate to each method. So that, looking at QC on a daily basis, nothing changes, the QC software flags results that fail the rules and we follow up.... Regardless of the QC protocol.



As is often the case when we try to take a good idea and use it in the real world, there are some practical challenges. To estimate the Sigma metric we need three values: Total Allowable Error, bias and the CV.

CV is relatively straight forward if we have QC samples that are targeted near the decision points of interest. We can use the CV from the QC material. We have to make sure that we are calculating the CV using enough data. 10 values is no where near enough and even 20 values will not give a robust estimate of CV. It is best to use data from several months of QC testing if possible.

We have already discussed the challenges with determining total allowable error so won't go over that again. However we recognize there is effort involved is choosing the best value to use.

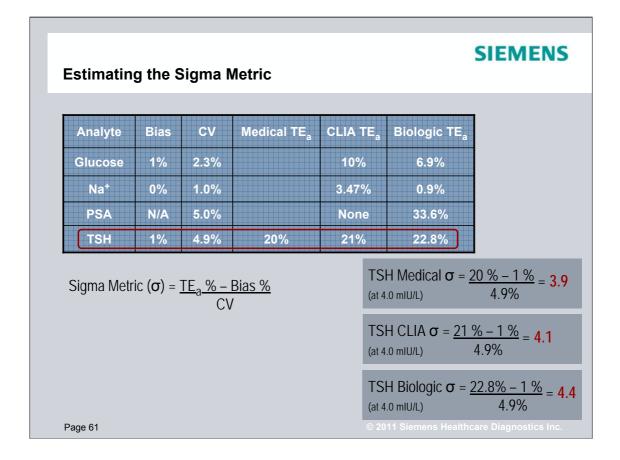


Finally there is bias. This can be a difficult challenge. Bias represents how much our results differ from the true result on the average. But what is truth ? How do we know what the true value is ? In articles about using sigma metric, it is often suggested that we use the QC or PT peer group mean as our measure of truth. But is that really the best choice? The peer group mean is not necessarily the most accurate value only the most popular one. It is entirely possible that the peer group is generally more biased than we are.

In the past folks have used the all method mean from PT results as truth, and at one time it may have given a reasonable estimate. However, today for many, many methods there is a predominant market leader that most labs are using and the all method mean is really nothing more than the peer group mean for that method. If that method is unbiased, then it is fine ... but how do we know that method is unbiased?

We can send samples to a reference or commercial lab to get comparative results. However, often these labs use the same methods we do. Sometimes however, these large labs do have reference methods, or something very close, available. If that's the case then those results could give us a good estimate of bias. What we really want is comparative results for fresh patient samples from a real reference method. Unfortunately that is almost impossible to find. Reference methods are very manual and are usually not practical for routine use. So we cannot afford to set them up and often cannot find a lab that can. In recent times some PT programs have begun assigning target values using reference type methods and grading is against the reference result rather than the peer group. If that's the case, those TP targets may be useful.

One pragmatic way to get started using the concept of sigma metric even if we cannot find a good way to estimate bias is to assume bias is zero. If we do this, we can estimate a sigma metric and use it to help set up out QC and generally we will get close to the ideal. Most methods do not have large biases so this can work at a very basic level to help us get started. Then once we find an estimate of bias that we feel accurately represents method bias with patient samples, we can revise our estimate of Sigma metric and adjust accordingly.



Looking at our example assays, there is only one, TSH, for which we have documented Error goals based on all three approaches medical use, CLIA limits and biologic data. Let's follow TSH through the process.

For the goal based on medical use we get a sigma metric of 3.9. Using the CLIA based goal we get a sigma metric of 4.1 and using the biologic goal we get a sigma metric of 4.4. All pretty much the same and all indicate that we can monitor and control TSH to meets these goals using standard statistical QC protocols.

However, that is not the case for all assays

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Analyte	Bias	с۷	CLIA TE <sub>a</sub>	Biologic TE <sub>a</sub>	CLIA σ	Biologic σ
Glucose	1%	2.3%	10%	6.9%	3.9	2.6
Na <sup>+</sup>	0%	1.0%	3.47%	0.9%	3.5	0.8
PSA	N/A	5.0%	None	33.6%	None	6.7
тѕн	1%	4.9%	21%	22.8%	4.1	4.4

## **Challenges:**

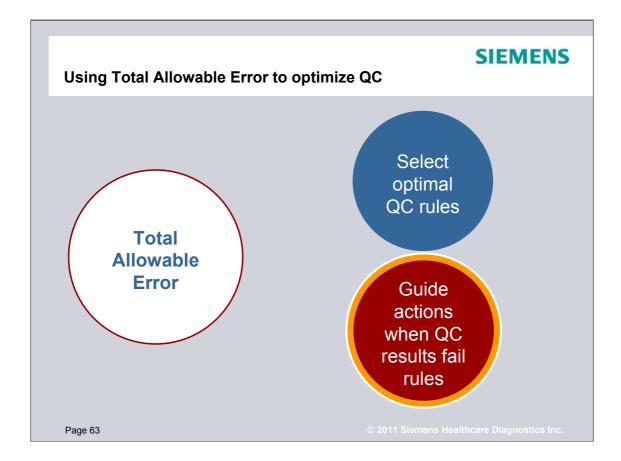
- · For some, no method in routine use has performance to meet biologically based goal
- For others, no medical or CLIA based performance goals are available
- There is no simple uniform way to set goals

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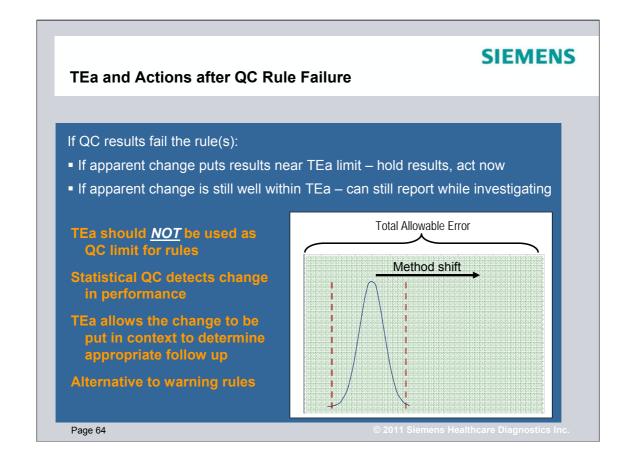
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When we look at our four example assays we see some of the challenges we face. For some assays the biologically based goals may not be achievable with current methods and technology. For other analytes, there may not be defined goals using criteria other than the biologic criteria. So we find that there is no simple uniform way to set Total Allowable Error goals and estimate the sigma metric. It becomes a decision based on available information and judgment.

However, it is worth the effort because it is so useful in helping us set up the most efficient and effective QC protocols.



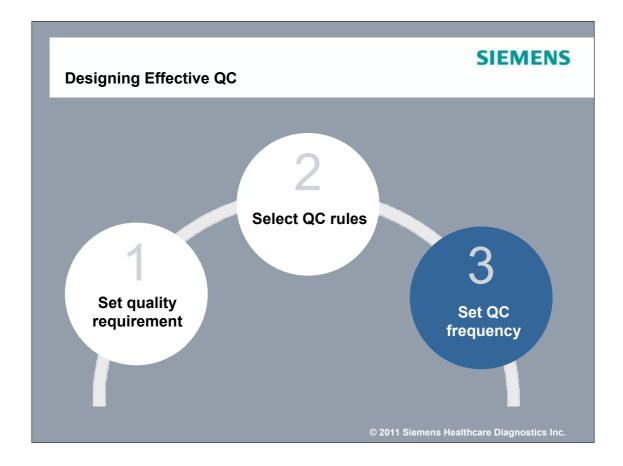
There is another way that Total Allowable Error can help us in looking at QC results and that is to guide our actions when we have a QC rule failure.



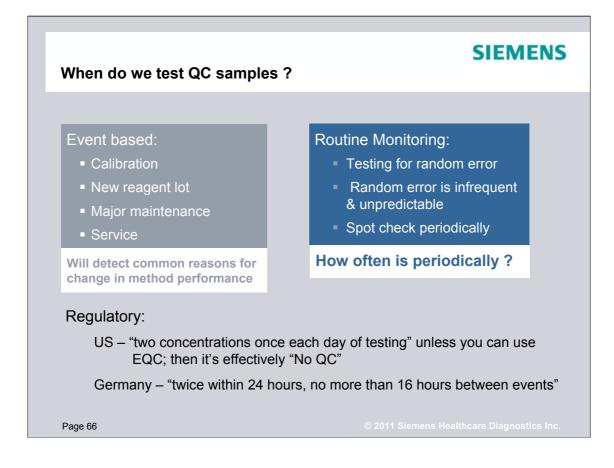
Our QC rules are statistically based and are designed to detect any change in method performance. If the apparent change in performance puts assay results near the limit of the Total Allowable Error, then all results should be held until the investigation is complete and the issue resolved.

However, if the shift in performance cause a QC rule failure, but the results are still comfortably within the Total Allowable Error limit, then results can still be reported while the investigation is being done. This is because in spite of the change in method performance the error in the results still is not large enough to affect medical decisions.

Some points to note ... This does NOT mean that we should use total allowable error limits as the acceptable limits for our QC rules. That would not work very well at all. We want our QC rules to work for us to detect any change in method performance. Then we can use total allowable error to put this change in context relative to medical decision making. Once we have put the method performance in context, we can make the appropriate decisions about how to proceed and whether patient results can be released. In this regard Total allowable error can function like a warning rule.

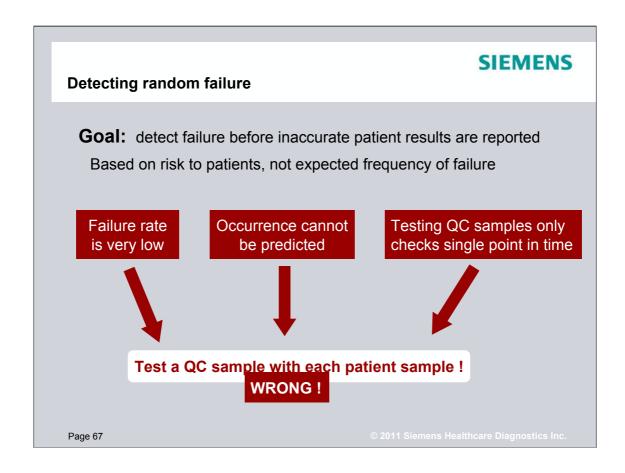


So we have set our quality requirement and used it to help select the optimal QC rules, now we need to establish when to test QC samples in order to finalize our QC protocol



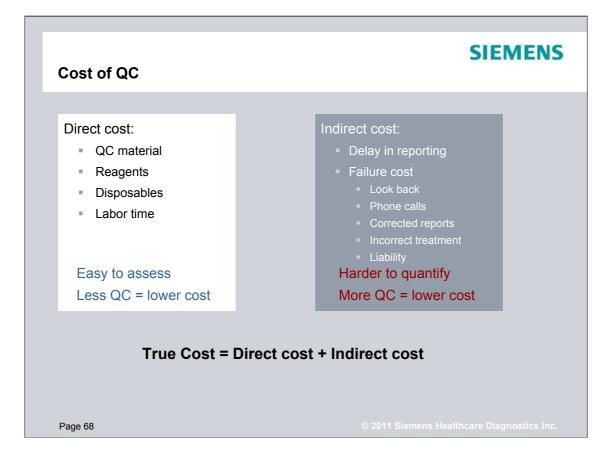
When do we test QC samples? Generally there are two triggers for QC testing. One is event based. We test QC samples every time we do something that may have altered the performance of the system. Things like calibrate, maintenance, new reagent lots, etc. The second trigger is based on routine monitoring to detect random error. We know any analytical system can fail. We know these failures are random in nature and infrequent. So we cannot predict when they will occur. As a consequence we periodically test QC samples as a spot check for this random error. But how often is periodically?

Even regulatory agencies cannot agree. In the US CLIA says the MINIMUM is once every 24 hours of testing. In Germany, the requirement is twice in 24 hours with no more than 16 hours between events. So how do we decide ?



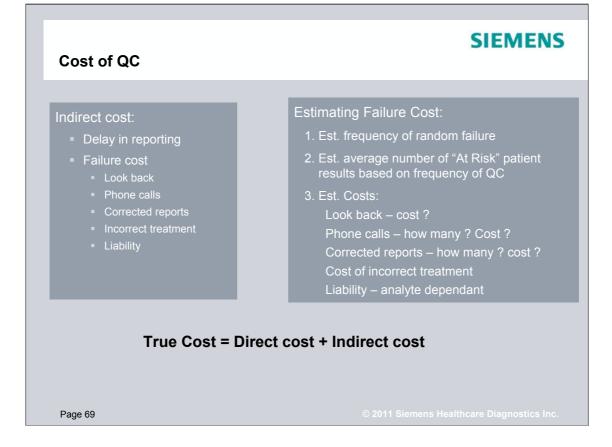
Our goal is to detect failure before any incorrect patient results are reported. So our goal is really risk based. We are more concerned about how many patient results might be incorrect than we are about how often the system might fail.

We know the failure rate is low. We cannot predict when the failure will occur. We know that testing QC samples can only tell us how the system is performing at the moment the QC sample is tested. So the obvious conclusion is to test a QC sample with every patient sample just to be sure ! WRONG !!! Clearly this conclusion is not workable. It is completely impractical because of the realities of workflow and the associated costs. So what do we do. We have to balance our need to reduce patient risk with the practical realities and costs. Let's look at cost in more detail



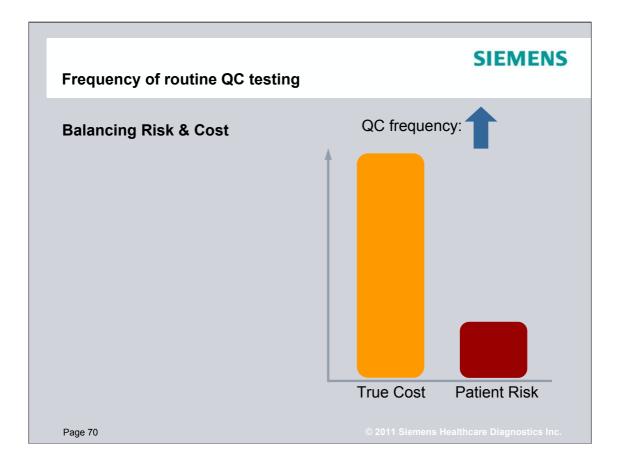
The direct costs of QC are fairly easy to understand and estimate. They include the cost of the QC material, the reagents and disposables used and the labor cost. But there is another cost to QC – the indirect costs. The costs resulting from delayed reporting of patient results because we are running QC samples on the instrument and investigating all the false positive QC rule failures before we report results. There's also the failure cost. This is the costs associated with the occurrence of a QC rule failure that is then determined to be a true failure. The costs of any look backs at patient results. The direct costs of any repeat testing of patient samples. The cost of phone calls and corrected reports. The costs of incorrect treatment decisions because of incorrect labs results and the potential liability costs of the incorrect results. Fortunately these last two are not often a big concern because few treatment decisions are made solely on the basis of a single lab result. However, it can happen.

To understand the true cost of what ever QC protocol we use, we have to estimate the indirect costs and factor that into the total cost. Direct costs are easy to assess and generally, the less QC we do, the lower the direct cost. Indirect costs are tougher to estimate and generally the less often we test QC sample, the greater the potential indirect costs.

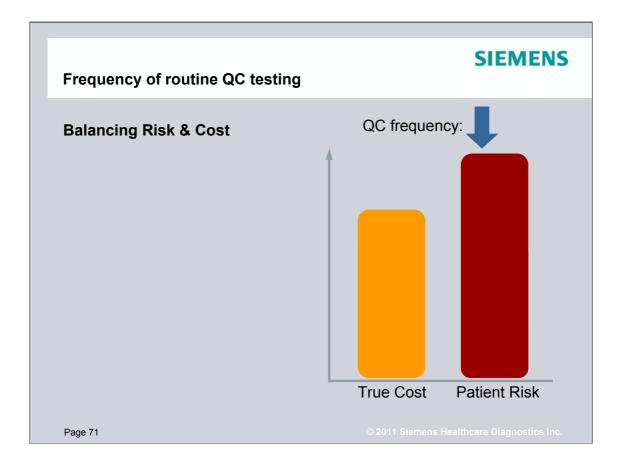


Let's look at estimating failure cost. First we need to estimate how often a real failure of the system is likely to occur. This will be fairly infrequent. Remember, the common reasons for changes to system performance are all event based and we are addressing them with our event based QC. Our concern here is the random failure. Then we need to estimate how many patient results are at risk if a failure occurs. Generally the average number of patient results at risk is half the number of results that would likely be reported between any teo routine QC events. Now we look at the costs of following up on those at risk patient results. Based on the lab's protocol, what is the look back process ? How many patient samples are retested, if any ? What is the likelihood of phone calls and corrected reports and estimate the cost. Then we have to factor in some cost for the possibility of incorrect treatment or liability. While an event like this may have huge costs, it will be a rare occurrence, so the cost we factor in can be modest.

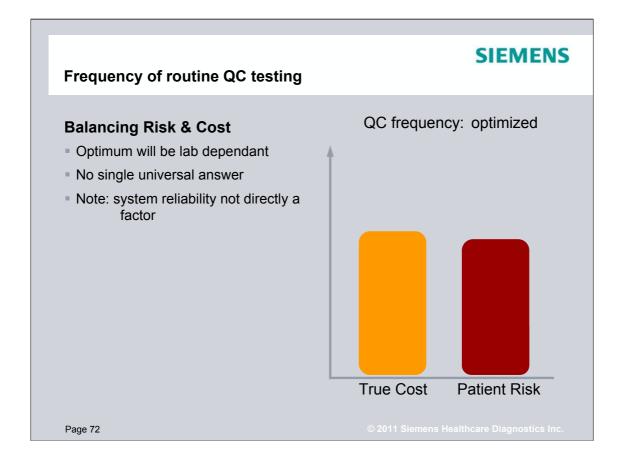
Now our true costs is the sum of the direct costs plus the indirect costs and we can play "what if" by looking at varying the frequency of routine QC testing and see what happens to the over all true cost. Lower QC frequency lowers direct and increases indirect. So with a little experimenting using our own testing volumes and protocols we can get an idea how to minimize the true cost.



In the end we try to balance the true cost of our QC protocol with patient risk. If we increase the frequency of QC testing, we lower patient risk, but our costs go up.

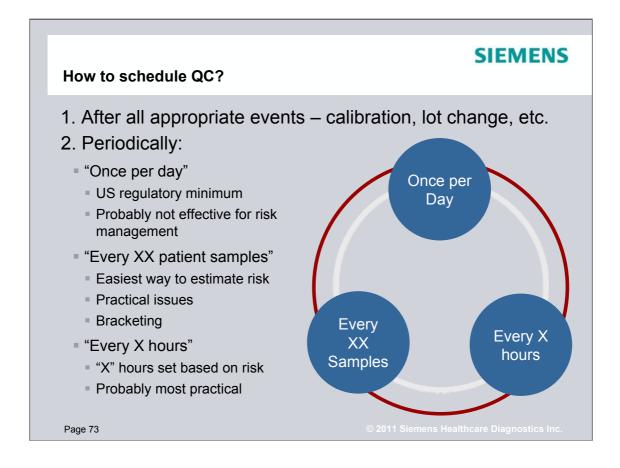


If we decrease QC frequency, we increase patient risk, but our costs go down. However, the costs don't drop as much as we might expect. Decreasing QC frequency lowers direct cost, but increases indirect cost.



We also recognize that we can never eliminate patient risk no matter how often we test QC samples. So the optimal protocol balances risk and cost and tries to get the most benefit in risk reduction for a true cost that can be sustained.

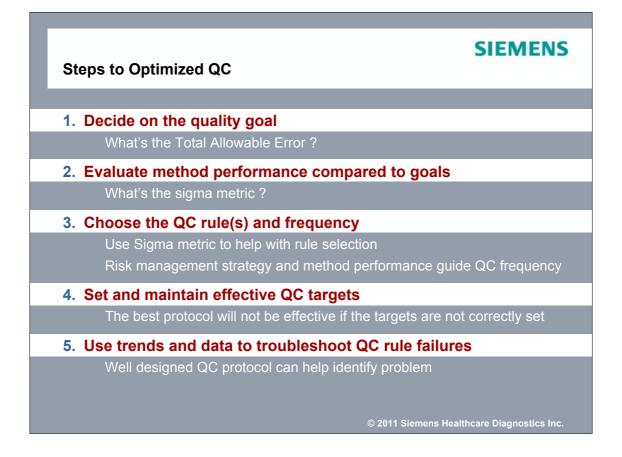
This optimum will be different for each laboratory. There is no single universally correct answer. We each have to figure it out. Also note that in this discussion, the expected frequency of system failure was not a factor used. That is because once the expected failure rate drops below a threshold, the risk management aspect of the QC protocol becomes more important than the expected frequency of failure.



So we have decided based on true cost and risk management how often we may want to test QC samples. Now how do we implement that? First, we test QC samples after every event that may alter system performance. Then for the periodic testing, what are the options. The CLIA minimum of once per day is probably not adequate to effectively manage indirect costs and patient risk for most laboratories. Remember, just because we are doing something that is the legal minimum, that doesn't mean we are doing it the best way possible.

Another way we can schedule QC samples is every XX patient samples. This makes it very easy to estimate how many patient samples may be at risk if we have a true failure, but it can be an awkward way to schedule QC. Since testing volume varies widely between analytes, this approach can have us testing QC samples for small groups of different methods quite often. This has a negative impact on workflow and drives up direct costs. This approach is also difficult to use unless QC testing can be auto-scheduled by the instrument, middleware or LIS. Folks working on the instrument cannot possibly keep track of how many samples have been tested for a given method. This approach is the foundation of QC bracketing, which is used in some labs and is mandated for some testing.

Finally there is the way most of us schedule routine QC... every X hours. Using the approach we have discussed we would use our estimates balancing total cost and risk to decide how long a time we should have between each QC event. This is probably the most practical approach because we can keep track of the time interval manually and increasingly instruments, middleware, etc. can autoschedule QC based on time. If we use the approach we have discussed to determine the optimum time interval, this can be an effective way to do QC.



Let's review the steps to optimize our QC protocol.

1. We decide on our quality goal - and use total allowable error to help

2. We compare actual method performance to the quality goals and calculate the sigma metric

3. We use the sigma metric to help select the optimal QC rules and the true costs and risk management to determine QC frequency

4. We set and maintain effective QC targets

5 We use the tools built into our QC protocol to help us troubleshoot when we do find a problem

This approach can help us use QC to best advantage assure the highest possible quality while still having a program that is practical and cost effective.