LEARNING OBJECTIVES:

2.03.01 Identify five general types of errors that can occur when analyzing radioactive samples, and describe the effect of each source of error on sample measurements.

2.03.02 State two applications of counting statistics in sample analysis.

2.03.03 Define the following terms:
   a. mean
   b. median
   c. mode

2.03.04 Given a series of data, determine the mean, median, or mode.

2.03.05 Define the following terms:
   a. variance
   b. standard deviation

2.03.06 Given the formula and a set of data calculate the standard deviation.

2.03.07 State the purpose of a Chi-squared test.

2.03.08 State the criteria for acceptable Chi-squared values at your site.

2.03.09 State the purpose of creating quality control (QC) charts.

2.03.10 State the requirements for maintenance and review of QC charts at your site.

2.03.11 State the purpose of calculating warning and control limits.

2.03.12 State the purpose of determining efficiencies and correction factors.

2.03.13 Given counting data and source assay information, calculate efficiencies and correction factors.

2.03.14 State the meaning of counting data reported as $x \pm y$. 

LEARNING OBJECTIVES: (continued)

2.03.15 Given counting results and appropriate formulas, report results to desired confidence level.

2.03.16 State the purpose of determining background.

2.03.17 State the method and requirements for determining background for counting systems at your site.

2.03.18 State the purpose of performing sample planchet maintenance.

2.03.19 State the method and requirements for performing planchet maintenance for counting systems at your site.

2.03.20 Explain methods to improve the statistical validity of sample measurements.

2.03.21 Define "detection limit" and explain the purpose of using detection limits in the analysis of radioactive samples.

2.03.22 Given the formula and necessary information, calculate detection limit values for counting systems at your site.

2.03.23 State the purpose and method of determining crosstalk.

2.03.24 State criteria for acceptable values of crosstalk for counting systems at your site.

2.03.25 State the purpose of performing a voltage plateau.

2.03.26 State the method of performing a voltage plateau on counting systems at your site.
INTRODUCTION

Radiological sample analysis involves observation of a random process, one that may or may not occur, and estimation of the amount of radioactive material present based on that observation. All over the country radiation protection personnel are using activity measurements to make decisions that may affect the health and safety of workers at those facilities and their surrounding environments.

This unit presents an overview of measurement processes, and statistical evaluation of both measurements and equipment performance. In addition, this unit addresses some of the actions to take to minimize the sources of error in count room operations.

GENERAL SOURCES OF ERROR

Assuming the counting system is calibrated correctly, there are five general sources of error associated with counting a sample.

1. **Self-absorption**
2. **Backscatter**
3. **Resolving time**
4. **Geometry**
5. **Random disintegration** of radioactive atoms (statistical variations).

**Self-Absorption**

Self-absorption is the absorption of the emitted radiation by the sample material. Self-absorption could occur for:

- Liquid samples with a high solid content
- Air samples from a high dust area
- Use of the improper filter paper may introduce a type of self-absorption, especially
in alpha counting (absorption by the media, or filter).

When a sample has an abnormally high amount of material on the sample media, it could introduce an additional error due to self-absorption. Personnel performing samples should ensure the correct sample media is used and that the sample does not become too heavily loaded with sample. Count room personnel should be checking samples for improper media or heavily loaded samples.

**Backscatter**

Backscatter occurs when the emitted radiation traveling away from the detector is reflected, or scattered back, by the material in back of the sample. The amount of radiation that is scattered back into the detector will depend on the type and energy of the radiation and the type of backing material (reflector). The amount of backscattered radiation increases as the energy of the radiation increases and as the atomic number of the backing material increases. Generally, the backscatter error is only a consideration for particulate radiation, such as alpha and beta particles. Because beta particles are more penetrating than alpha particles, the backscatter error will be more pronounced for beta particles. The ratio of measured activity of a beta source counted with a reflector compared to counting the same source without a reflector is called the backscatter factor (BF).

(Equation 1)

\[
BF = \frac{\text{counts w/ reflector}}{\text{counts w/out reflector}}
\]

Normally, the backscatter error is accounted for in the efficiency or conversion factor of the instrument. However, if different reflector materials, such as aluminum and stainless steel are used in calibration and operation, an additional unaccounted error is introduced. This additional error will be about 6% for stainless steel versus aluminum. Count room personnel must be aware of the reflector material used during calibration of the counting equipment. Any deviation from that reflector material will introduce an unaccounted error and reduce confidence in the analysis results.

**Resolving Time**

Resolving time is the time interval which must elapse after a detector pulse is counted before another full-size pulse can be counted. Any radiation entering the detector during the resolving time will not be recorded as a full size pulse; therefore, the information on that radiation interaction is lost. As the activity, or
decay rate, of the samples increases, the amount of information lost during the resolving time of the detector is increased. As the losses from resolving time increase, an additional error in the measurement is introduced. Typical resolving time losses are shown in Table 1.

Table 1. Percent Loss in Various Detectors

<table>
<thead>
<tr>
<th>Count rate (cpm)</th>
<th>GM Tube$^1$</th>
<th>Proportional$^2$</th>
<th>Scintillation$^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>20,000</td>
<td>1.7%</td>
<td>&lt; 1%</td>
<td>&lt; 1%</td>
</tr>
<tr>
<td>40,000</td>
<td>3.3%</td>
<td>&lt; 1%</td>
<td>&lt; 1%</td>
</tr>
<tr>
<td>60,000</td>
<td>5.0%</td>
<td>&lt; 1%</td>
<td>&lt; 1%</td>
</tr>
<tr>
<td>100,000</td>
<td>8.3%</td>
<td>&lt; 1%</td>
<td>1.0%</td>
</tr>
<tr>
<td>300,000</td>
<td>25.0%</td>
<td>&lt; 1%</td>
<td>3.5%</td>
</tr>
<tr>
<td>500,000</td>
<td>42.0%</td>
<td>1.5%</td>
<td>5.8%</td>
</tr>
</tbody>
</table>

$^1$GM tube: 50µs  
$^2$Proportional detector: 2µs  
$^3$Scintillation detector: 7µs

Resolving time losses can be corrected by using the equation:

\[
R = \frac{R_o}{1 - R_o \tau}
\]

(Equation 2)

where: 
\(R\) = "true" count rate, in cpm  
\(R_o\) = observed count rate, in cpm  
\(\tau\) = resolving time of the detector, in minutes ("tau")

Count room personnel should be aware of the limitations for sample count rate based on procedures and the type detector in use to prevent the introduction of additional resolving time losses. This is especially true for counting equipment that uses GM detectors.
Geometry

Geometry of the sample and detector describes the positioning of the sample in relation to the detector. Radiation is emitted from a sample equally in all directions. Normally, only a fraction of the emitted radiation is emitted in the direction of the detector because the detector does not surround the sample. If the distance between the sample and the detector is varied, then the fraction of emitted radiation which hits the detector will change. This fraction will also change if the orientation of the sample under the detector (side-to-side) is varied.

An error in the measurement can be introduced if the geometry of the sample and detector is varied from the geometry used during instrument calibration. This is especially critical for alpha counting where any change in the sample to detector distance also increases or decreases the chance of shielding the alpha particles by the air between the sample and detector.

Examples of geometry-related errors are:

• Piling smears and/or filters on top of each other in the same sample holder. Piling of the samples moves the top sample closer to the detector and varies the calibration geometry.

• Using deeper or shallower sample holders than those used during calibration changes the sample-to-detector distance.

• Adjusting movable bases in the counting equipment sliding drawer changes the sample to detector distance.

• Using too many or not using the appropriate sample holder or planchet changes the sample to detector distance. Sources not fixed in position can change can change geometry and reduce reproducability.

• Plexiglass shelving in counting chamber is improperly set.

Random Disintegration

The fifth source of general counting error is the random disintegration of the radioactive atoms and constitutes the remainder of the lesson.

STATISTICS
Statistics is a branch of mathematics that deals with the organization, analysis, collection, and interpretation of statistical data. No definition of statistical data is given. However, Webster's does define a statistic as "an estimate of a variable, as an average or a mean, made on the basis of a sample taken from a larger set of data."

This last definition is applicable to our discussion of counting statistics. After all, when we take samples, we use the data derived from analysis of those samples to make determinations about conditions in an area, in water, or in air, etc., assuming that the sample is representative.

So, we have estimated conditions (a variable) on the basis of a sample (our smear, water sample, air sample) taken from a larger set of data.

Over the years, various methods and observations have identified three models which can be applied to observations of events that have two possible outcomes (binary processes). Luckily, we can define most observations in terms of two possible outcomes. For example, look at the following table:

<table>
<thead>
<tr>
<th>Trial</th>
<th>Definition of Success</th>
<th>Probability of Success</th>
</tr>
</thead>
<tbody>
<tr>
<td>Tossing a coin</td>
<td>&quot;heads&quot;</td>
<td>1/2</td>
</tr>
<tr>
<td>Rolling a die</td>
<td>&quot;a six&quot;</td>
<td>1/6</td>
</tr>
<tr>
<td>observing a given radioactive nucleus for a time, t</td>
<td>The nucleus decays during the observation</td>
<td>$1 - e^{-t}$</td>
</tr>
</tbody>
</table>

For each of the processes that we want to study, we have defined a trial (our test), a success and a failure (two possible outcomes) and have determined the probability of observing our defined success.

Now, to study these processes, we can use proven, statistical models to evaluate our observations for error. Consider the possibilities when throwing two dice. There are 36 possible outcomes when throwing two dice, as indicated in Table 3.
Table 3. Possibilities in Rolling Dice

<table>
<thead>
<tr>
<th>Result</th>
<th>Possibilities</th>
<th>No. of Possibilities</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>1&amp;1</td>
<td>1</td>
</tr>
<tr>
<td>3</td>
<td>1&amp;2, 2&amp;1</td>
<td>2</td>
</tr>
<tr>
<td>4</td>
<td>1&amp;3, 2&amp;2, 3&amp;1</td>
<td>3</td>
</tr>
<tr>
<td>5</td>
<td>1&amp;4, 2&amp;3, 3&amp;1, 1&amp;3&amp;2</td>
<td>4</td>
</tr>
<tr>
<td>6</td>
<td>1&amp;5, 2&amp;4, 3&amp;3, 4&amp;2, 5&amp;1</td>
<td>5</td>
</tr>
<tr>
<td>7</td>
<td>1&amp;6, 2&amp;5, 3&amp;4, 4&amp;3, 5&amp;2, 6&amp;1</td>
<td>6</td>
</tr>
<tr>
<td>8</td>
<td>2&amp;6, 3&amp;5, 4&amp;4, 5&amp;3, 6&amp;2</td>
<td>5</td>
</tr>
<tr>
<td>9</td>
<td>3&amp;6, 4&amp;5, 5&amp;4, 6&amp;3</td>
<td>4</td>
</tr>
<tr>
<td>10</td>
<td>4&amp;6, 5&amp;5, 6&amp;4</td>
<td>3</td>
</tr>
<tr>
<td>11</td>
<td>5&amp;6, 6&amp;5</td>
<td>2</td>
</tr>
<tr>
<td>12</td>
<td>6&amp;6</td>
<td>1</td>
</tr>
</tbody>
</table>

If, in our study of this process, we define a success as throwing a number between 2 and 12, the outcome is academic. All trials will be successful, and we can describe the probabilities of throwing any individual number between the range of 2 and 12 inclusive would add up to 1.

If we define a success as throwing a particular number, we can define the probability of our success in terms of the number of possible outcomes that would give us that number in comparison to the total number of possible outcomes.

If we graph the data presented in the table above, we have some additional options.
The area under the curve can be mathematically determined and would correspond to the probability of success of a particular measurement. For example, to determine the probability of throwing a number between 2 and 12 we would calculate the area under the curve between 2 and 12 and the results of that calculation would be 36.

If we were to take two dice, roll the dice a large number of times, and graph the results in the same manner, we would expect these results to produce a curve that is the same shape as the one discussed earlier.

This is what statistics is all about. Random binomial processes should produce results in certain patterns that have been proven over the years. The three models that are used are distribution functions of binomial processes with different governing parameters. These functions and their restrictions are:

- **Binomial Distribution**

  This is the most general of the statistical models and widely applicable to all processes with a constant probability. It is not widely used in nuclear applications because the mathematics are too complex.

- **Poisson Distribution**

  A simplified version of binomial distribution is the *Poisson* (pronounced "pwusówn") and is valid when the probability of success is small. The observation time (count time) is short in comparison to half life, or a very low detection efficiency. The number of decays that are observed is very small in comparison to the number of atoms present to decay.

- **Gaussian Distribution**

  Also called the "normal distribution," the *Gaussian* (pronounced "Gowziun") distribution is a further simplification which is applicable if the average number of successes is relatively large, but the probability of success is still low. This is applicable, since the probability of success is low in all radioactivity measurements. A sample containing more activity would, therefore, provide more successes.
2.03.02 State the two purposes for statistical analysis of count room operations.

**Application of Statistical Models**

Application of specific statistical methods and models to nuclear counting operations is termed *counting statistics* and is essentially used to do two things:

- **Predict the inherent statistical uncertainty associated with a single measurement**, thus allowing us to estimate the precision associated with that measurement.
- **Serve as a check** on the normal function of nuclear counting equipment.

2.03.03 Define the following terms:

a. mean
b. median
c. mode

**Definitions**

**Mode**  An individual data point that is repeated the most in a particular data set.

<table>
<thead>
<tr>
<th>Student</th>
<th>Test Score</th>
</tr>
</thead>
<tbody>
<tr>
<td>Susan</td>
<td>80</td>
</tr>
<tr>
<td>Richard</td>
<td>82</td>
</tr>
<tr>
<td>Greg</td>
<td>86</td>
</tr>
<tr>
<td>Peter</td>
<td>88</td>
</tr>
<tr>
<td>Andrew</td>
<td>90</td>
</tr>
<tr>
<td>Wanda</td>
<td>92</td>
</tr>
<tr>
<td>Randy</td>
<td>95</td>
</tr>
<tr>
<td>Jennifer</td>
<td>95</td>
</tr>
<tr>
<td>Sarah</td>
<td>95</td>
</tr>
</tbody>
</table>

*Figure 2. Sample Data Set*
For example, in the set of test scores above, a score of 95 occurs (i.e. is repeated) more often than any other score.

**Median**
The center value in a data set arranged in ascending order.

For example, in the same set of test scores as above, this is the score in the middle where one half the scores are below and the other half are above the median. The median for the above set is 90.

**Mean**
Average value of all the values in the data set. This is simply found by adding all of the values together and dividing by number of values in the set. For the set above the mean would be 89.2.

### Determination of mode, median, and mean

**Determination of the mode** of a particular data set requires no real calculation, just observation. In the data set containing 1.8, 1.8, 1.9, 1.9, 2.0, 2.0, 2.0, 2.0, 2.2, 2.3, 2.3, 2.3, 2.5, observation reveals that 2.0 appears more often than any other measurement, thus, 2.0 is the mode.

**Determination of the median** first requires arrangement of individual data points into ascending order. The median is then the centerpoint. In the data set used above, both 7th and 8th data points are 2.0, thus the median is 2.0. If the 7th and 8th data points were different, the median would be the average of the two values.

**Determination of the mean** requires that all data points in the data set be added and the sum divided by the number of data points in the set. In the set used above, the sum is 29 and there were 14 data points, thus the mean is 2.07.

Mean determination is often expressed using special symbols, as illustrated in the following equation:

\[
\bar{x} = \frac{\sum x_i}{n}
\]

(Equation 3)
where: \( x_i \) = data point with index \( i \)
\( \bar{x} \) = mean (sometimes pronounced "x bar")
\( n \) = number of data points
\( \Sigma \) = summation symbol \( \sum_{i=1}^{n} x_i = x_1 + x_2 + x_3 + \cdots + x_n \)

**Poisson Distribution**

Remember that the parameters included a low probability of success, \( P(x) \), and number of successes, \( x \), is low. If we graphed a Poisson distribution function, we would expect to see the predicted number of successes at the lower end of the curve with successes over the entire range if sufficient trials were attempted. Thus, the curve would appear as seen in Figure 3.

![Poisson Distribution](image)

**Figure 3. Predicted Successes for Poisson**

The Poisson model is used mainly for applications involving system background and detection limits, where the population (number of counts) is small. This will be discussed in greater detail later.

**Gaussian distribution**
In this distribution, we said that the predicted number of successes is large. A graph of a Gaussian distribution function appears in Figure 4.

![Gaussian Distribution](image)

**Figure 4. Predicted Successes for Gaussian**

Note that the highest number of successes is at the center of the curve, the curve is a bell shaped curve, and the relative change in success from one point to the adjacent is small. Also note that the mean, or average number of successes, is at the highest point, or at the center of the curve.

The Gaussian, or normal distribution, is applied to counting applications where the mean success is expected to be greater than 20. It is used for calibrations and operational checks, as well as for normal samples containing activity. It may or may not include environmental samples (very low activity).

2.03.05 Define the following terms:
   a. variance
   b. standard deviation

2.03.06 Given the formula and a set of data calculate the standard deviation.

Using the Gaussian distribution model depicted in Figure 5 we need to define
several descriptive terms and measures used in statistical calculations.

The amount of scatter of data points around the mean is defined as the sample variance. In other words, it tells how much the data "varies" from the mean. A more precise term is the standard deviation, represented by \( \sigma \) (pronounced "sigma"). Mathematically, in the normal distribution, the standard deviation is the square root of the variance.

The variance and standard deviation represent the spread of the population, or, the data set, in a normal distribution. The distribution of the population is thus easier to interpret by use of the standard deviation.

**Standard Deviation**

The population standard deviation is defined mathematically as:

\[
\sigma = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n}}
\]

(Equation 4)

where: \( \sigma \) = biased population standard deviation  
\( x_i \) = sample counts for each data point
\[ \bar{x} = \text{mean} \]
\[ n = \text{number of data points} \]

If most of the data points are located close to the mean, the curve will be tall and steep and have a low numerical value for a standard deviation. If data points are scattered, the curve will be lower and not as steep and have a larger numerical value for a standard deviation.

In a Gaussian distribution, it has been determined mathematically that 68.2\% of the area under the curve falls within the data point located at the mean ± (plus or minus) one standard deviation (1\(\sigma\)); 95.4\% of the area under the curve falls between the data point located at ± two standard deviations (2\(\sigma\)), etc. What this means to us in terms of counting processes is that if the distribution (as depicted in Figure 5) is representative of a counting function with a mean observable success greater than twenty (Gaussian distribution) is that for certain percentages of the observations the observed successes will be within correlating numbers of standard deviations, as shown in Table 4.

<table>
<thead>
<tr>
<th>Percentage of Observations</th>
<th>Area under Curve (Standard Deviations)</th>
</tr>
</thead>
<tbody>
<tr>
<td>68.2%</td>
<td>± 1(\sigma)</td>
</tr>
<tr>
<td>95.4%</td>
<td>± 2(\sigma)</td>
</tr>
<tr>
<td>99.97%</td>
<td>± 3(\sigma)</td>
</tr>
</tbody>
</table>

Remember, the area of the curve represents the probability of success in a random process. In radiological protection this random process is the decay of a radioactive sample. We will see shortly how the statistical distribution can be utilized.

Practically speaking, the known statistical distribution is used in radiological protection when setting up a counting system and in evaluating its operation by means of daily pre-operational checks. In performing the calibration of the system a radioactive source with a known activity is counted twenty times for one minute each time. Using the data from the twenty counts the mean and standard deviation can be calculated. The mean can be used to determine the efficiency of the system while allowing for a certain number of standard deviations during operation. The twenty counts can also be used to perform another required test of the system's
Calculate the mean and sample standard deviation for the following data set: 
{151, 161, 143, 145, 121, 150, 135, 142, 135, 146}.

Using a table or spreadsheet we can arrange the data points and determine the required statistical values:

<table>
<thead>
<tr>
<th>n</th>
<th>x</th>
<th>(x - x)²</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>151</td>
<td>65.61</td>
</tr>
<tr>
<td>2</td>
<td>161</td>
<td>327.61</td>
</tr>
<tr>
<td>3</td>
<td>143</td>
<td>0.01</td>
</tr>
<tr>
<td>4</td>
<td>145</td>
<td>4.41</td>
</tr>
<tr>
<td>5</td>
<td>121</td>
<td>479.61</td>
</tr>
<tr>
<td>6</td>
<td>150</td>
<td>50.41</td>
</tr>
<tr>
<td>7</td>
<td>135</td>
<td>62.41</td>
</tr>
<tr>
<td>8</td>
<td>142</td>
<td>0.81</td>
</tr>
<tr>
<td>9</td>
<td>135</td>
<td>62.41</td>
</tr>
<tr>
<td>10</td>
<td>146</td>
<td>9.61</td>
</tr>
</tbody>
</table>

Therefore, the mean of the set is found to be 142.9, and the standard deviation is 10.31.
State the purpose of a Chi-squared test.

Chi-squared Test

The Chi-squared test (pronounced "ki") is used to determine the precision of a counting system. Precision is a measure of how exactly a result is determined without regard to its accuracy. It is a measure of the reproducibility of a result, or in other words, how often that result can be repeated, or how often a success can be obtained.

The Chi-squared test is often referred to as a "goodness-of-fit" test. It answers the question: How well (or "good") does this data fit a normal distribution curve? If it does NOT fit a curve indicating sufficient randomness, then the counting instrument may be malfunctioning.

This test results in a numerical value called the Chi-squared value, $X^2$, which is then compared to a range of values for a specified number of observations. This range represents the expected (predicted) probability for the chosen distribution. If the $X^2$ value is lower than the expected range, this tells us that there is not a sufficient degree of randomness in the observed data. If the value is too high, it tells us that there is too much randomness in the observed data.

The Chi-squared value is calculated as follows:

$X^2 = \frac{\sum (x_i - \bar{x})^2}{\bar{x}}$

(Equation 5)

Example 2.

Using the data from Example 1, determine the Chi-squared value for the data set.

$X^2 = \frac{1062.9}{142.9} = 7.44$

The twenty 1-minute counts of a source mentioned previously are used to perform the Chi-squared test. The criteria for the test will be discussed shortly.

The Chi-squared test is used to ensure that the data obtained meets the model of a
normal distribution. This means that the probability (range) of values occurring within ± 2 standard deviations of the mean is already defined by the normal distribution. In fact, the values to which the Chi-squared value is compared represent the -2σ and +2σ variations from the mean in either direction, that is, 5% and 95% probability respectively for the area under the curve.

The Chi-squared values that correlate to the associated probability for a set of twenty counts are:
2.03.08  State the criteria for acceptable Chi-squared values at your site.

(Insert site-specific information here.)

Assuming the data passes the Chi-squared test, the data can be used to prepare quality control charts for use on a daily basis to verify the consistent performance of the system.

2.03.09  State the purpose of creating quality control (QC) charts.

2.03.10  State the requirements for maintenance and review of QC charts at your site.

QUALITY CONTROL CHARTS

Quality control charts are prepared using source counting data obtained during system calibration. The source used for daily checks should be identical to, if not the very same as, the one used during system calibration. Obviously since this test verifies that the equipment is still operating within an expected range of response, we cannot change the conditions of the test in mid-stream. QC charts, then, enable us to track the performance of the system while in use.

Quality control charts should be maintained in the area of the radioactivity counting system such that they will be readily accessible to those who operate the system. It is incumbent upon operators to review them prior to using the system so as to ensure that the charts are current.

Data that can be used includes gross counts, counts per unit time and efficiency. Most nuclear laboratories use a set counting time corresponding to the normal counting time for the sample geometry being tested. If smears are counted for one minute, then all statistical analysis should be based on one-minute counts. The easiest data to use would be a daily one-minute source count.

When the system is calibrated and the initial calculations performed, the numerical values of the mean ± 1, 2 and 3 standard deviations are also determined.

Using standard graph paper, paper designed specifically for this purpose, or a computer graphing software, lines are drawn all the way across the paper at those points corresponding to the mean,
the mean plus 1, 2, and 3 standard deviations: and the mean minus 1, 2, and 3 standard deviations. The mean is the center line of the paper.

2.03.11 State the purpose of calculating warning and control limits.

System Operating Limits

The values corresponding to ±2 and ±3 standard deviations are called the upper and lower warning and control limits, respectively. The results of the daily source counts are graphed daily. Most of the time our results will lie between the lines corresponding to ±1 standard deviation (68.2%). We also know that 95.4% of the time our count will be between ±2 standard deviations and that 99.97% of the time our count will be between ±3 standard deviations.

Counts that fall outside the warning limit (±2σ) are not necessarily incorrect. Statistical distribution models say that we should get some counts in that area. Counts outside the warning limits indicate that something MAY be wrong. The same models say that we will also get some outside the control limits (±3σ). However, not very many measurements will be outside those limits. We use 3σ as the control a standard for acceptable performance. In doing so we say that values outside of ±3σ indicate unacceptable performance, even though those values may be statistically valid.

True randomness also requires that there be no patterns in the data that is obtained; some will be higher than the mean, some will be lower, and some will be right on the mean.

When patterns do show up in quality control charts, they are usually indicators of systematic error. For example:

- Multiple points outside two sigma
- Repetitive points (n out of n) outside one sigma
- Multiple points, in a row, on the same side of the mean
- Multiple points, in a row, going up or down.

The assumption is made that systematic error is present in our measurements, and that our statistical analysis has some potential for identifying its presence. However, industry assumption is that systematic error that is present is very small in comparison to random error.
COUNTER EFFICIENCY

A detector intercepts and registers only a fraction of the particles emitted by a radioactive source. The major factors determining the fraction of particles emitted by a source that actuate a detector include:

- The fraction of particles emitted by the source which travel in the direction of the detector window
- The fraction emitted in the direction of the detector window which actually reach the window
- The fraction of particles incident on the window which actually pass through the window and produce an ionization
- The fraction scattered into the detector window

All radiation detectors will, in principle, produce an output pulse for each particle or photon which interacts within its active volume. The detector then would be said to be 100 percent "efficient," because 100 percent of the activity was detected and reported. In practice, because of the factors outlined above, the actual (total) activity emitted from the source is not detected. Therefore, there is only a certain fraction of the disintegrations occurring that results in counts reported by the detector. Using a calibrated source with a known activity a precise figure can be determined for this fraction. This value can then be used as a ratio in order to relate the number of pulses counted to the number of particles and/or photons incident on the detector. This ratio is called the efficiency. It can also be referred to as the detector yield, since the detector yields a certain percentage of the actual counts.

The detector efficiency gives us the fraction of counts detected per disintegration, or c/d. Since activity is the number of disintegrations per unit time, and the number of counts are detected in a finite time, the two rates can be used to determine the efficiency if both rates are in the same units of time. Counts per minute (cpm) and disintegrations per minute (dpm) are the most common.
Thus, the efficiency, $E$, can be determined as shown in Equation 6. Used in this manner the time units will cancel resulting in counts/dis.

(Equation 6) \[ E = \frac{cpm}{dpm} = \frac{c}{d} \]

The efficiency obtained in the formula above will be in fractional (decimal) form. To calculate the percent efficiency, the fraction can be multiplied by 100. For example, an efficiency of 0.25 would mean $0.25 \times 100$, or 25%.

**Example 3.**

A source is counted and yields 2840 counts per minute. If the source activity is known to be 12,500 dpm calculate the efficiency and percent efficiency.

\[
E = \frac{2840}{12500} = 0.2272 \\
0.2272 \times 100 = 22.72\%
\]

By algebraic manipulation, Equation 6 can be solved for the disintegration rate (see Equation 7). The system efficiency is determined as part of the calibration. When analyzing samples a count rate is reported by the counting system. Using Equation 7 the activity, $A$, of the sample can then be determined in dpm, and then converted to any other units of activity (e.g. Ci, Bq).

(Equation 7) \[ dpm = \frac{cpm}{E} \quad \rightarrow \quad A_{dpm} = \frac{cpm}{E} \]
Example 4.

A sample is counted on a system with a 30% efficiency. If the detector reports 4325 net counts per minute what is the activity of the sample in dpm?

\[
A = \frac{4325}{0.3} = 14416.7 \text{ dpm}
\]

As seen in Equation 7 above, the net count rate is divided by the efficiency. A correction factor (CF), which is simply the inverse of the efficiency, is used by multiplying it times the net count rate to determine the activity, as in Equation 8.

(Equation 8)

\[
CF = \frac{1}{E}
\]

Example 5.

An instrument has an efficiency of 18%. What is the correction factor?

\[
CF = \frac{1}{0.18} = 5.5
\]

This count-rate correction factor should not to confused with a geometry correction factor used with some radiation instruments, such as the beta correction factor for a Cutie Pie (RO-3C).
2.03.14 State the meaning of counting data reported as \( x.xx \pm yy \).

2.03.15 Given counting results and appropriate formulas, report results to desired confidence level.

ERROR CALCULATIONS

The error present in a measurement governed by a statistical model can be calculated using known parameters of that model. Nuclear laboratories are expected to operate at a high degree of precision and accuracy. However, since we know that there is some error in our measurements, we are tasked with reporting measurements to outside agencies in a format that identifies that potential error. The format that is used should specify the activity units and a range in which the number must fall. In other words, the results would be reported as a given activity plus or minus the error in the measurement. Since nuclear laboratories prefer to be right more than they are wrong, counting results are usually reported in a range that would be correct 95\% of the time, or at a 95\% confidence level.

In order to do this, the reported result should be in the format:

\[
x.xx \pm yy (K\sigma)
\]

where:
- \( x.xx \) = activity, in units of dpm, Ci, or Bq
- \( yy \) = associated potential error in the activity
- \( K \) = multiple of counting error
- \( \sigma \) = standard deviation at stated confidence level (CL)

Note: Use of \( K\sigma \) is only required for confidence levels other than 68\% (see Table 6). Therefore:

\[
\begin{align*}
\sigma &= 1 \times \sigma \quad 68\% \text{ CL} \quad \text{(optional)} \\
1.64\sigma &= 1.64 \times \sigma \quad 90\% \text{ CL} \quad \text{(sometimes used)} \\
2\sigma &= 1.96 \times \sigma \quad 95\% \text{ CL} \quad \text{(normally used)}
\end{align*}
\]

For example, a measurement of \( 150 \pm 34 \text{ dpm} \) indicates the activity as 150 dpm; however, it could be as little as 116 dpm or as much as 184 dpm with 95\% confidence (2\( \sigma \)). In practice the error in the measurement is first determined in the resulting count rate. Then the counting system efficiency is applied to the result and to the associated error, as well as any required conversion factors in order to report the measurement in the appropriate units associated with the applicable limits.

The calculations of the actual range of error is based on the standard definition for the
distribution. In the normal or Gaussian distribution, the a single count is the mean, or \( x = \bar{x} \). Additionally, the standard deviation of a single count is defined as the square root of the mean, or \( \sigma = \sqrt{x} \). The error, \( e \), present in a single count is some multiplier, \( K \), multiplied by the square root of that mean, i.e. some multiple times the standard deviation, \( K\sigma \). The value of \( K \) used is based on the confidence level that is desired, and is derived from the area of the curve included at that confidence level (see Figure 5). Common values for \( K \) include:

<table>
<thead>
<tr>
<th>Error</th>
<th>Confidence Level</th>
<th>( K )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Probable</td>
<td>50%</td>
<td>0.6745</td>
</tr>
<tr>
<td>Standard</td>
<td>68%</td>
<td>1.0000</td>
</tr>
<tr>
<td>9/10</td>
<td>90%</td>
<td>1.6449</td>
</tr>
<tr>
<td>95/100</td>
<td>95%</td>
<td>1.9600</td>
</tr>
<tr>
<td>99/100</td>
<td>99%</td>
<td>2.5750</td>
</tr>
</tbody>
</table>

To calculate the range to the point at which you would expect to be right 95% of the time, you would multiply the standard deviation times 1.96 and report the results of the measurement as \( x.xx \text{ dpm} \pm yy \text{ dpm} (2\sigma) \). Note that using a 68% or 50% confidence level introduces an expected error a large percentage of the time. Therefore, for reasonable accuracy a higher confidence level must be used.

The simple standard deviation (\( \sigma \)) of the single count (\( x \)) is usually determined as a count rate (counts per unit time). This is done by dividing the count rate (\( R \)) by the count time (\( T \)). Subscripts can be applied to distinguish sample count rates from background count rates.

\[(Equation 10)\]
\[
\sigma = K \sqrt{\frac{R}{T}}
\]
Example 6.

The count rate for a sample was 250 cpm. Assume 10 minute counting time, zero background counts and a 25% efficiency. Report sample activity at a 95% C.L.

\[
2\sigma = 1.96 \sqrt{\frac{250}{10}} = 1.96 \sqrt{25}
\]
\[
2\sigma = 1.96(5) = 9.8
\]
\[
\frac{250 \text{ cpm}}{0.25 \text{ c/d}} = 1000 \text{ dpm}
\]
\[
\frac{9.8 \text{ cpm}}{0.25 \text{ c/d}} = 39 \text{ dpm}
\]

Therefore, the sample activity should be reported as:

\[
1000 \pm 39 \text{ dpm (2}\sigma)\]

2.03.16 State the purpose of determining background.

BACKGROUND

Determination of Background

Radioactivity measurements cannot be made without consideration of the background. Background, or background radiation, is the radiation that enters the detector concurrently with the radiation emitted from the sample being analyzed. This radiation can be from natural sources, either external to the detector, such as cosmic or terrestrial, or radiation originating inside the detector chamber that is not part of the sample.

In practice the total counts are recorded by the counter. This total includes the counts contributed by both the sample and the background. Therefore, the contribution of the background will produce an error in measurements of radioactivity unless the background count rate is determined by a separate operation and subtracted from the total activity, or gross count rate. The difference between the gross and the background rates is called the net count rate (sometimes given units of ccppm, or corrected counts per minute). This
relationship is seen in the following equation:

(Equation 11) \[ R_S = R_{S+B} - R_B \]

where:
- \( R_S \) = net sample count rate (cpm)
- \( R_{S+B} \) = gross sample count rate (cpm)
- \( R_B \) = background count rate (cpm)

The background is determined as part of the system calibration by counting a background (empty) planchet for a given time. The background count rate is determined in the same way as any count rate, where the the gross counts are divided by the count time, as seen in Equation 12 below.

(Equation 12) \[ R_B = \frac{N_B}{T_B} \]

where:
- \( R_B \) = background count rate (counts per time, i.e. cpm)
- \( N_B \) = gross counts, background
- \( T_B \) = background count time

For low-background counting systems two background values must be determined; one for alpha and one for beta-gamma. These two values are used to determine alpha and beta-gamma sample count rates respectively during calibration and when analyzing samples.

In practice, background values should be kept as low as possible. As a guideline, background on automatic counting systems should not be allowed to exceed 0.5 cpm alpha and 1 cpm beta-gamma. If system background is above this limit the detector should be cleaned or replaced.

**Reducing Background**

As will be discussed shortly, the lower the system background the more reliable the analysis of samples will be. In low-background counting systems the detector housing is surrounded by lead shielding so as to reduce the background. Nonetheless, some background still manages to reach the detector. Obviously, little can be done to reduce the actual source of background due to natural sources. On many systems a second detector is incorporated which detects
background penetrating radiation. When a sample is analyzed the counts detected by this second detector during the same time period are internally subtracted from the gross counts for the sample.

Background originating inside the detector chamber can be, for the most part, more easily controlled. The main contributors of this type of background are:

- Radiation emitted from detector materials
- Radioactive material on inside detector surfaces
- Radioactive material on the sample slide assembly
- Contamination in or on the sample planchet or planchet carrier

There are, unfortunately, trace amounts of radioactive isotopes in the materials of which the detector and its housing are made. This is simply a fact of life in the atomic age. However, the contribution to background from this source is negligible, but should nonetheless be acknowledged.

Radioactive material can be transferred from contaminated samples to the inside surfaces of the detector chamber during counting. This usually occurs when samples having gross amounts of material on them are counted on a low-background system. In the insertion and withdrawal of the sample into the detector chamber loose material can be spread into the chamber. In order to prevent this these samples should be counted using a field survey instrument or a mini-scaler. Low-background systems are designed for counting lower-activity samples. Counting of a high-activity sample on these systems should be avoided unless it is a sealed radioactive source.

Radioactive material can also be transferred from contaminated samples to the slide assembly upon which samples are inserted into and withdrawn from the detector chamber. This can be prevented in the same way as stated above. In addition, when loading and stacking samples for counting, ensure that the slide assembly cover is in place. The slide assembly should also be cleaned on a routine basis, at least weekly.

When loading and unloading samples into and from planchets material from the samples can be spread to the planchet and/or the carrier plate. Most smears and air samples are 47-mm diameter and are counted in a planchet that is almost the same size. The planchet is placed in a carrier which surrounds and supports the planchet and allows for automatic sample exchange by the counting system. When
a sample is counted the entire carrier is placed under the detector window inside the detector chamber. Any contamination from previous samples on the carrier or in the planchet is counted with and attributed to the sample.

A paper disc can be placed in the bottom of the planchet as a step in preventing transfer of material from samples to the planchet. Care should be taken when loading and unloading samples such that material remains on the sample media.

2.03.17 State the method and requirements for determining background for counting systems at your site.

(Insert site-specific information here.)

2.03.18 State the purpose of performing planchet maintenance.

**Planchet Maintenance**

Planchets and carriers should be inspected, cleaned and counted on a routine basis. All in-use planchets and carriers must read less than established site limits. Planchets exceeding these limits should be decontaminated and recounted as necessary.

By maintaining planchets clean and as free from contamination as possible sample result reliability will be increased because the amount of error introduced in the sample analysis will be reduced.

2.03.19 State the method and requirements for performing planchet maintenance for counting systems at your site.
PROPAGATION OF ERROR

The error present in a measurement includes the error present in the sample count, which contains both sample and background, and the error present in the background count. Rules for propagation of error preclude merely adding the two errors together. The total error in the measurement can be calculated by squaring the error in the background and adding that to the square of the error in the sample count and taking the square root of the sum, as shown in Equation 13.

(Equation 13)  \[ e_S = \sqrt{e_{S+B}^2 + e_B^2} \]

where:  
- \( e_S \) = error present in the measurement (sample)  
- \( e_{S+B} \) = error in sample count (sample plus background)  
- \( e_B \) = error present in background count

Since we normally use this equation in terms of a count rate, the formula can be slightly modified as follows, and the error stated as the sample standard deviation (\( \sigma \)):

(Equation 14)  \[ K\sigma_s = K \sqrt{\frac{R_{S+B}}{T_s} + \frac{R_B}{T_B}} \]

where:  
- \( R_{S+B} \) = gross sample count rate (sample plus background)  
- \( R_B \) = background count rate  
- \( T_s \) = sample count time  
- \( T_B \) = background count time  
- \( K \) = confidence level multiple (see Table 6)

The error in the sample count is the standard deviation of the count, which is the square root of that count (see Equation 13 above). If we square a square root we get the number we started with.

Example 7.
An air sample is counted and yields 3500 counts for a 2-minute count period. The system background is 10 cpm determined over a 50-minute count time. Determine the error in the sample and report the net count rate to 95% confidence level.

\[
\frac{3500 \text{ counts}}{2 \text{ minutes}} = 1750 - 10 = 1740 \text{ cpm}
\]

\[
\sigma_j = \sqrt{\frac{1750}{2} + \frac{10}{50}} = 29.6
\]

\[
2\sigma = 29.6 \times 1.96 = 58
\]

Therefore, the net count rate should be reported as:

\[1740 \pm 58 \text{ cpm (2}\sigma)\]

For the case where the sample counting time and the background counting time is the same, the formula can be simplified even more to:

\[
(Equation\ 15)\quad K\sigma_j = K\sqrt{\frac{R_{S-B} + R_B}{T}}
\]
Example 8.

A long-lived sample is counted for one minute and gives a total of 562 counts. A one minute background gives 62 counts. Report net sample count rate to 95% CL.

\[ R_s = 562 - 62 = 500 \text{ cpm} \]

\[
2\sigma_s = 1.96 \sqrt{\frac{562 + 62}{1}}
\]

\[
2\sigma_s = 1.96 \sqrt{624}
\]

\[
2\sigma_s = 1.96(24.98)
\]

\[
2\sigma = 49
\]

Therefore, the net sample count rate and associated error is:

\[ 500 \pm 49 \text{ cpm (2\sigma)} \]

2.03.20 Explain the methods used to improve the statistical validity of count room measurements.

IMPROVING STATISTICAL VALIDITY OF COUNT ROOM MEASUREMENTS

Minimizing the statistical error present in a single sample count is limited to several options. If we look at the factors present in the calculation below (same as Equation 14), we can see that there are varying degrees of control over these factors. The standard deviation is calculated here in terms of count rate.

\[
\sigma_{rate} = \sqrt{\frac{R_s + B}{T_s} + \frac{B}{T_B}}
\]

\( R_{s+B} \) is the sample count rate. We really have no control over this.

\( R_B \) is the background count rate. We do have some control over this. On any counting equipment the background should be maintained as low as possible. In most of our counting applications, however, the relative magnitude of the background count rate should be extremely small in comparison to the sample count rate if proper procedures are followed. This really becomes an issue when counting samples for free release or environmental samples. However,
some reduction in error can be obtained by increasing the background counting time, as discussed below.

$T_B$ and $T_S$ are the background and sample counting times, respectively. These are the factors that we have absolute control over. In the previous section we talked about how reliable the count itself is. We have been able to state that a count under given circumstances may be reproduced with a certain confidence level and that the larger the number of counts the greater the reliability. The condition we have been assuming is that our count is taken within a given time. So, in order to get more precise results, many counts must be observed. This requires, then, that if we have low count rates, the counting time must be increased in order to obtain many counts, thereby making the result more precise.

To obtain this precision for a sample with a low count rate, the total counting time depends upon both the sample and background count rates. We can see that to get precise results the sample and background counts must be taken for comparable times. For high sample activities the sample count time can be relatively short compared to the background count time. For medium count rates we must increase the sample count time in order to increase precision.

As the sample activity gets even lower, we approach the case where we must devote equal time to the background and source counts. In other words, by counting low activity samples for the same amount of time as that of the background determination, we increase the precision of our sample result. However, we must never count a sample for a period of time longer than that of the system background.

In summary, by minimizing the potential error present, we improve statistical validity of our measurements.
DETECTION LIMITS

The detection limit of a measurement system refers to the statistically determined quantity of radioactive material or radiation that can be measured or detected at a preselected confidence level. This limit is a factor of both the instrumentation and technique or procedure being used.

The two parameters of interest for a detector system with a background response greater than zero are (see Figure 6):

- \( L_c \)  
  Critical detection level: the response level at which the detector output can be considered "above background"

- \( L_n \)  
  Minimum significant activity level, i.e. the activity level that can be seen with a detector with a fixed level of certainty

These detection levels can be calculated by the use of Poisson statistics, assuming random errors and systematic errors are separately accounted for, and that there is a background response. For these calculations, two types of statistical counting errors must be considered quantitatively in order to define acceptable probabilities for each type of error:

- **Type I** - occurs when a detector response is considered above background when in fact it is not (associated with \( L_c \))

- **Type II** - occurs when a detector response is considered to be background when in fact it is greater than background (associated with \( L_n \))
Figure 6. Errors in Detection Sensitivity

If the two probabilities (areas labeled I and II in Figure 6) are assumed to be equal, and the background of the counting system is not well-known, then the critical detection level ($L_C$) and the minimum significant activity level ($L_D$) can be calculated. The two values would be derived respectively using the equations $L_C = k \sigma_B$ and $L_D = k^2 + 2k \sigma_B$. If 5% false positives (Type I error) and 5% false negatives (Type II error) are selected to be acceptable levels, i.e. 95% confidence level, then $k = 1.645$ and the two equations can be written as:

(Equation 16)

$$L_C = 1.645 \sqrt{\frac{R_B}{T_B} + \frac{R_B}{T_S}}$$

(Equation 17)

$$L_D = 2.71 + 3.29 \sqrt{\frac{R_B}{T_B} + \frac{R_B}{T_S}}$$
where: $L_C = \text{Critical detection level}$  
$L_D = \text{a priori detection limit [minimum significant activity level}$  
$k = \text{Poisson probability sum for I and II (assuming I and II}$  
probabilities are equal)$  
$R_B = \text{background counts}$  
$T = \text{count time (sample and background)}$

The minimum significant activity level, $L_D$, is the *a priori* (before the fact) activity level that an instrument can be expected to detect 95% of the time. In other words, it is the smallest amount of activity that can be detected at a 95% confidence level. When stating the detection capability of an instrument, this value should be used.

The critical detection level, $L_C$, is the lower bound on the 95% detection interval defined for $L_D$ and is the level at which there is a 5% chance of calling a background value “greater than background.” This value ($L_C$) should be used when actually counting samples or making direct radiation measurements. Any response above this level should be counted as positive and reported as valid data. This will ensure 95% detection capability for $L_D$.

If the sample count time ($T_s$) is the same as the background count time ($T_B$), then equations 16 and 17 can be simplified as follows:

(Equation 18) \[ L_C = 2.32 \sqrt{\frac{R_B}{T}} \]

(Equation 19) \[ L_D = 2.71 + 4.65 \sqrt{\frac{R_B}{T}} \]
$L_D = \text{Minimum significant activity level (count rate)}$
$k = \text{same as above; 1.645 for 95\% CL}$
$R_B = \text{background count rate}$
$T_B = \text{background count time}$
$T_S = \text{sample count time}$

Therefore, the full equations for $L_C$ and $L_D$ must be used for samples with count times differing from the background determination time (95\% CL used). These equations assume that the standard deviation of the sample planchet-carrier background during the sample count (the planchet-carrier assumed to be 0 activity) is equal to the standard deviation of the system background (determined using the background planchet-carrier).

The critical detection level, $L_C$, is used when reporting survey results. It is used to say that to a 95\% confidence level, samples above this value are radioactive. This presupposes, then, that 5\% of the time clean samples will be considered radioactive.

The minimum significant activity level, $L_D$, [referred to as the LLD (Lower Limit of Detection) in some texts] is calculated prior to counting samples. This value is used to determine minimum count times based on release limits and airborne radioactivity levels. In using this value we are saying that to a 95\% CL samples counted for at least the minimum count time calculated using the $L_D$ that are positive will be radioactive (above the $L_C$). This presupposes, then, that 5\% of the time samples considered clean will be radioactive.

**Example 9.**

A background planchet is counted for 50 minutes and yields 16 counts. Calculate the critical detection level and the minimum significant activity level for a 0.5 minute sample count time.

\[
L_C = 1.645 \sqrt{\frac{0.32}{50} + \frac{0.32}{0.5}} \\
L_C = 1.645 \sqrt{0.0064 + 0.64} \\
L_C = 1.645 \sqrt{0.6464} \\
L_C = 1.32 \text{ cpm}
\]

\[
L_D = 2.71 + 3.29 \sqrt{\frac{0.32}{50} + \frac{0.32}{0.5}} \\
L_D = 2.71 + 3.29(0.804) \\
L_D = 5.36 \text{ cpm}
\]
Given the formula and necessary information, calculate detection limit values for counting systems at your site.

(State the purpose and method of determining crosstalk.)

**CROSSTALK**

**Discrimination**

*Crosstalk* is a phenomenon that occurs on proportional counting systems (such as a Tennelec) that employ electronic, pulse-height discrimination, thereby allowing the simultaneous analysis for alpha and beta-gamma activity. Discrimination is accomplished by establishing two thresholds or *windows*. These windows can be set in accordance with the radiation energies of the isotopes of concern. Recall that the pulses generated by alpha radiation will be much larger than those generated by beta or gamma. This makes the discrimination between alpha and beta-gamma possible. Beta and gamma events are difficult to distinguish; hence, they are considered as one and the same type.

In practice the lower window is set such that electronic noise and ultra-low-energy photon events are filtered out. Any pulse generated whose size is greater than the setting for the lower window is considered an event, or a *count*. The upper window is then set such that any pulses which surpass the upper discriminator setting will be considered an alpha count (see Figure 7).

For output purposes the system routes each count to a series of *channels* which simply keep a total of the counts routed to them. Channel A is for alpha counts, Channel B is beta-gamma counts, and Channel C is total counts. As a sample is being counted all valid counts registered (i.e. those which surpass the lower discriminator setting) are routed to the C-channel. In addition, if the count was considered an alpha count (i.e. it surpassed the upper discriminator setting) it is routed to the A-channel; else it is tallied in the B-channel. In effect, as far as the number of counts reported for each channel, what occurs is the number of beta-
gamma counts (Channel B) are determined by subtracting the number of alpha counts (Channel A) from the total counts (Channel C). In other words, \( C - A = B \).

* \( C - A = B \)

**Figure 7. Pulse-height Discrimination**
Origin of Crosstalk

Now that we understand the process involved, there is a dilemma that stems from the fact that events are identified by the system as either alpha or beta-gamma according to the size of the pulse generated inside the detector. The system cannot really tell what type of radiation has generated the pulse. Rather, the pulse is labeled as "alpha" or "beta-gamma" by comparing the size of the pulse to the discriminator setting. It is the setting of the discriminator that poses the dilemma.

Alpha particles entering the detector chamber generally are attenuated by the detector fill-gas because of their high LET, thereby producing a large pulse. Low-energy beta particles and photons also will lose all their energy within the detector gas, but nevertheless produce a smaller pulse because of their lower energies. High energy beta particles can still retain some of their energy even after having produced a pulse while traversing the detector volume. Rather than leaving the detector, as would a photon, the beta is reflected off of the detector wall and reenters the volume of gas, causing ionizations and generating a second pulse. These two pulses can be so close together that the detector sees them as one large pulse. Because of the large pulse size it can surpass the upper discriminator setting and is, therefore, counted as an alpha, and not as a beta. This is crosstalk. The result is that alpha activity can be reported for a sample when in fact there was little or no alpha present. Conversely, if a true alpha-generated pulse is not large enough so as to exceed the upper discriminator, it would be counted as a beta-gamma event.

The solution is not a simple one. The setting of the upper discriminator depends on the radiations and energies of the sources and samples being analyzed. If high energy beta radiations are involved a significant portion of them could be counted as alpha events if the setting is too low. If the setting is too high, lower-energy alpha events could be counted as beta-gamma. In practice the setting of the discriminator will usually be a "happy medium." A discussion of how this can be dealt with is in order.

Calibration Sources and Crosstalk

For calibrations of Tennelec counting systems, Oxford Instruments, the manufacturer, provides the following general recommendations for discriminator settings. First, using a Strontium-90 beta source set the upper (α) discriminator such that there is 1% beta-to-alpha crosstalk. Then using a Polonium-210 source set the α+β discriminator such that there is less than 3% alpha-to-beta crosstalk.

Energies of sources used to calibrate counting systems should be the same as, or as
close to as possible, the energies of isotopes in the samples analyzed. Wherever possible they should be a pure emitter of the radiation of concern.

For beta-gamma sources the most popular isotope in radiological protection is Sr-90. It has a relatively long half-life of 29.1 years, but emits betas of only 546 keV. However, Sr-90 decays to Ytrium-90, another beta emitter which has a short half-life of only 2.67 days and emits a 2.281 MeV beta. Y-90 decays to Zirconium-90m which emits a 2.186 Mev gamma almost instantaneously to become stable. The daughters reach equilibrium with the strontium parent within a number of hours after source assay. Hence, for every Sr beta emitted a Y beta is also emitted, thereby doubling the activity. These sources are often listed as Sr/Y-90 for obvious reasons. This makes Sr/Y-90 sources an excellent choice and are used by many sites for calibrations and performance testing.

Po-210 is essentially a pure alpha emitter. This is primarily the reason why it is recommended. It yields a strong alpha, but it also has a short half-life. A comparison of some alpha emitters is given in Table 8.

### Table 8. Alpha Emitters

<table>
<thead>
<tr>
<th>Isotope</th>
<th>Half-Life</th>
<th>Energy (MeV)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Po-210</td>
<td>138.38 days</td>
<td>5.3044</td>
</tr>
<tr>
<td>Pu-239</td>
<td>2.4E4 years</td>
<td>5.156, 5.143, 5.105</td>
</tr>
<tr>
<td>Ra-226</td>
<td>1.60E3 years</td>
<td>4.78, 4.602</td>
</tr>
<tr>
<td>Th-230</td>
<td>7.54E4 years</td>
<td>4.688, 4.621</td>
</tr>
<tr>
<td>Natural U</td>
<td>4.4E9 years (avg.)</td>
<td>4.2 (avg.)</td>
</tr>
</tbody>
</table>

(Insert site-specific information here.)
VOLTAGE PLATEAUS

Very simply put, a plateau is a graph that indicates a detector's response to an isotope with variations of high voltage. The x-axis represents the high voltage and the y-axis is the counts. The resulting curve gives an indication of detector quality, and can indicate problems with the counting gas should they be present. The curve can be used to determine the optimum operating high voltage for the system.

Most automatic low-background counting systems provide several different analysis modes. These modes count samples at certain pre-determined voltages. Counting systems generally provide three analysis modes:

- ALPHA ONLY
- ALPHA THEN BETA
- ALPHA AND BETA (SIMULTANEOUS)

There are usually two voltage settings used in conjunction with these analysis modes:

- **Alpha** voltage (lower)
- **[Alpha plus] Beta** voltage (higher)

Recall that in a proportional counter the amount of voltage determines the amount of gas multiplication. Because of the high LET of alpha radiation, at a lower voltage, even though the gas amplification will be lower, alpha pulses will still surpass the lower discriminator and some will even pass the upper discriminator. Because of the lower gas amplification beta-gamma pulses will not be large enough to be seen. Therefore, any counts reported for the sample will be alpha counts.

In ALPHA ONLY mode the sample is counted once, at the alpha voltage. Counts may appear in either the A or B channels. Upon output the A and B channels will be added together and placed in Channel A and, therefore, reported as alpha counts; the B channel will be cleared to zero, thereby resulting in no beta-gamma counts.

In ALPHA THEN BETA mode the sample is counted twice. The first count interval determines the alpha counts using the alpha voltage. The second count is done at the beta voltage. The determination of alpha and beta-gamma counts in this mode is based strictly on the operating characteristics of the detector at the different voltages. For this reason, the A and B counts are
summed during both counting intervals to attain the total counts. The separation of alpha and beta-gamma counts is then calculated and reported according to the following formula:
2.03.26 State the method of performing a voltage plateau on counting systems at your site.

(Insert site-specific information here.)

In conjunction with initial system setup and calibration by the vendor two voltages plateaus are
performed—alpha voltage and beta voltage. For P-10 gas the alpha plateau is started at about 400 volts and the beta plateau at about 900 volts. Alpha and beta plateaus are defined by the isotope being used and not by the channel being used to accumulate the counts. More appropriately, the gross counts are accumulated and plotted for each type of isotope. Each time that a count is completed, the high voltage is incremented a specific amount, typically 25 to 50 volts, and another count is accumulated. This is repeated until the end of the range is reached, typically about 1800 volts.

With the high voltage set at the starting point, few or no counts are observed because of insufficient ion production within the detector. As the voltage is increased, a greater number of pulses are produced with sufficient amplitude to exceed the discriminator threshold, and are then accumulated in the counter. There will be a high voltage setting where the increase in counts levels off (see Figure 8). This area is the detector plateau. Further increases in high voltage result in little change in the overall count rate. The plateau should remain flat for at least 200 volts using a Sr/Y-90 source, and this indicates the plateau length. Between 1750 and 1850 volts the count rate will start to increase dramatically. This is the avalanche region, and the high voltage should not be increased any further.

**Beta Voltage Plateau**

Sr/Y-90 - Tennelec LB5100 #4

![Diagram of Beta Voltage Plateau](image)

**Figure 8. Voltage Plateau using Sr/Y-90**

The region where the counts level off is called the *knee* of the plateau. The operating voltage is...
chosen by viewing the plateau curve and selecting a point 50 to 75 volts above the knee and where the slope per 100 volts is less than 2.5%. This ensures that minor changes in high voltage will have negligible effects on the sample count. Poor counting gas or separation of the methane and argon in P-10 can result in a very high slope of the plateau.

Upon initial system setup and calibration the vendor determines and sets the optimum operating voltages for the system. Thereafter, plateaus should be generated each time the counting gas is changed.

SUMMARY

This lesson addressed the measures used to minimize error and the fundamentals of binomial statistics, as well as the application of these fundamentals in a nuclear counting environment. Completion of the unit does not qualify the student to perform any tasks independently.

REFERENCES:


5. Moe, Harold, Operational Health Physics Training; ANL-88-26; DOE; Argonne National Laboratory; Chicago; 1988.

