



Although I'm sure that some of you have these rules memorized from previous CLU-IN events, let's run through them quickly for our new participants.

Please mute your phone lines during the seminar to minimize disruption and background noise. If you do not have a mute button, press \*6 to mute #6 to unmute your lines at anytime. Also, please do NOT put this call on hold as this may bring delightful, but unwanted background music over the lines and interupt the seminar.

You should note that throughout the seminar, we will ask for your feedback. You do not need to wait for Q&A breaks to ask questions or provide comments. To submit comments/questions and report technical problems, please use the ? Icon at the top of your screen. You can move forward/backward in the slides by using the single arrow buttons (left moves back 1 slide, right moves advances 1 slide). The double arrowed buttons will take you to 1<sup>st</sup> and last slides respectively. You may also advance to any slide using the numbered links that appear on the left side of your screen. The button with a house icon will take you back to main seminar page which displays our agenda, speaker information, links to the slides and additional resources. Lastly, the button with a computer disc can be used to download and save today's presentation materials.

With that, please move to slide 3.













If the entire DU could be analyzed in a single giant analysis, there would be no uncertainty about the true Pb concentration. Note that this process would produce a result that represents a giant composite of all soil particles in the DU.



For this thought experiment, assume that there is no analytical error.

Since a single DU cannot be analyzed in a single analysis, we must take samples, analyze them, and then draw conclusions about the DU concentration from the concentration of the samples. In scenario A, we take 4 discrete (grab) samples. Because we want to use those samples to determine the actual concentration for the entire DU, we take the average of the 4 data points. Since we are using 4 small samples taken from a heterogeneous medium (soil), there is uncertainty in whether the average of the 4 data points accurately represents the concentration for the DU.

In scenario B, we take more samples (n = 33), but it is expensive to analyze them all. So we perform a physical averaging by combining all the samples (now called increments) together to form a single composite called an incremental sample, which is analyzed. This is equivalent to taking 33 samples and analyzing all individually, then mathematically averaging all 33 results. Again, because this is not a complete analysis, there is uncertainty about how close the sample average is to the true concentration.

In scenario C, we imagine splitting the entire DU into individual analytical samples. Each of those samples is analyzed individually so that the entire mass of the DU has been analyzed. If the mass of an analytical sample is 1 gram, and the mass of soil in the DU is 1 ton, then there are 1.4 million samples analyzed. The results from those samples are analyzed to determine the true concentration of the DU. Since the entire mass was analyzed, there is no uncertainty in the concentration calculated for the DU.

Obviously, scenario C is impossible, so we are left with designs that look like A or B. Which one looks like it would be more representative of the true concentration of the DU?



Because we cannot know the true concentration, we have to consider the possibility that our calculated mean is greater than the true concentration or less than the true. We also don't know by how much the over- or under-estimate occurs. To be protective, EPA is more concerned with the possibility of underestimating the true concentration.



In the face of uncertainty, being protective means that underestimates of the mean must be minimized. So a "buffer" is added to the calculated mean to compensate for the uncertainty in the estimate, with the hope that the new value will equal or exceed the true mean a certain % of the time. This sum of the "buffer" and the calculated mean is called an upper confidence limit. A 95% upper confidence limit (or UCL) is expected to equal or exceed the true mean 95% of the time. The width of the "buffer" is determined by 4 things: see next slide.



The width of the "buffer" is determined by 4 things: the type of statistical distribution the data come from, the number of samples in the data set used to estimate the mean, and the amount of variability in that data set, and the degree of confidence desired that the true concentration is not being underestimated.

The type of statistical distribution from which the data came determine which UCL formula is used. To use the Student's t UCL equation, the data should come from a normal (bell-shaped) or near-normal distribution.





• Distributions that conform to a mathematically defined curve are called parametric distributions.

• Nonparametric UCLs are also termed "distribution free" since the equation can be used with any data set. But if the data set does fit a parametric distribution, it is better to use the UCL equation specific to that distribution, because nonparametric UCLs are almost always higher than parametric UCLs.











SU results DO NOT represent a mean over the DU, but rather the concentration of some smaller area within the DU.

## Calculating a Mean & UCL for a DU Composed of <u>Unequal</u> SUs

- If the SUs are NOT all of the same area/volume, then weighting according to the relative area/volume must be done
- Cannot use ProUCL (doesn't do weighted calculations)
- Mean, SD and degrees of freedom all must be weighted to calculate the UCL
- Weighted mean & SD used to calculate the Chebyshev UCL
- More info in ITRC ISM1 doc, Section 4.4.1
- Weighted DU Calculator can be downloaded from webinar Resources list (Excel file)

2/24/2012

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No associated notes.



![](_page_22_Figure_0.jpeg)

![](_page_23_Figure_0.jpeg)

The goal is to figure out how many incremental samples to take, how many increments to take per incremental sample, and how many subsampling replicates should be analyzed per incremental sample.

The pilot study mean is assumed to reasonably estimate the true mean.

The subsampling SD and between-increment SD come from the Variability QC Procedure discussed on Day 2.

![](_page_24_Figure_0.jpeg)

VSP module: If you want to stay with 3 ISM replicates, but want a statistically indicated number of increments, sequentially enter values for the number of increments (1<sup>st</sup> brown circle) until number of incremental samples ("MI samples") goes to 4 (blue boxes).

True Average vs. Fixed Threshold	
Average vs. Fixed Threshold Sample Placement Costs Analytes	
I can 💌 assume the data will be normally distributed. For Help, highlight	an item
I want to use multiple increment v sampling. Static These design parameters apply to Analyte 1 Specify Null Hypothesis: I want to assume the site is unacceptable (dirty) v until proven otherwise. (Assume the true mean >= action level.) Specify False Rejection Rate (alpha) and Action Level: I want at least 95.0 % confidence that I will conclude the site is unacceptable (dirty) if the true mean is at or above the action level of 5000 units. Specify Width of Gray Region (delta) and False Accentance Rate (beta):	If the mean from the pilot study had been 4000, instead of 2660, then the width between the mean and action level would only be 1000. This narrower width causes the number of samples to go up, because it is harder statistically to tell the difference between 4000 and 5000 than it is
If the true means 1000 units below the action level (that is, 4000 units)	between 2660 and 5000.
over 1 wate no more than a 10.0       % chance or incorrectly accepting the hull         hypothesis that the site is unacceptable (true mean >= action level).         Specify Multiple Increment Sampling Options:         The estimated standard deviation between increments is 2975         Units.         The estimated standard deviation between analytical subsamples is 526         Units.         The number of increments in each Multiple Increment (MI) sample         is fixed         and the number of analytical subsamples taken from each MI sample         For this design I want to require that each MI sample will consist of 30         Increments	The statistics predict that if using 30 increments per ICS, you would need 7 ICSs to meet statistical goals.
Minimum Number of MI Samples for Analyte 1: 7	r Module 3.2 26

True Average vs. Fixed Threshold							
verage vs. Fixed Threshold Sample Placement Costs Analytes Total Area to Sample: 5000.00 Feet^2 Sampling Costs Fixed Planning and Validation Cost: \$ 1000.00 Per Increment Collection and Combination Cost: \$ 15.00 × 210 Increments Cost per Analytical Subsample Measurement: \$ 400.00 × 7 Analyses	Can enter costs & the module will optimize the number of increments & ICSs based on price of increment collection vs. price of analysis						
Optimal Sampling	VSP will also plot out sampling designs with coordinates						
Option         * 0 min         * 0 min <th< th=""><th>△ ◊ △ ○</th></th<>	△ ◊ △ ○						
4 7 28 1 1000 \$6.740.00 5 9 16 1 1000 \$6.760.00 Curret 7 30 1 1000 \$6.50.00	o 0 Δ •						
Select Cost Comparison Option:  Select Select If 1 analysis is performed on each multiple increment sample composed of 30	• △ ○ ◊						
increments, then 7 of these multiple increment samples are required to achieve 95% confidence that the true site mean is below the action level.	27						

![](_page_27_Figure_0.jpeg)

![](_page_28_Figure_0.jpeg)

Focus on identification and investigation of known or suspected spill areas as separate DUs.

This is the best you can do.

Collecting a grid of discrete samples and removing individual sample locations as "hot spots"

is erroneous and gives a false sense of understanding about the site.

In addition isn't our fundamental question, "What is the average concentration for a given exposure area = volume of soil?"

What volume does a discrete sample represent? – nothing more than the sample size you collected.

For a ISM sample, the volume represents the decision unit, therefore the analytical result represents the DU/SU.

It's not about dilution, it is about the volume over which you wish to know the average value of a parameter.

Photograph from Ryan et al, 2004, Reducing Children's Risk From Lead in Soil: Environmental Sciences & Technology, January 1, 2004.

An example of incremental versus discrete data for the same property is coming up.

![](_page_29_Figure_0.jpeg)

With a 175 ft grid spacing, a circular hotspot needs at least a 100-ft radius (200 ft diameter or  $\frac{3}{4}$  acre) to be detected 95% of the time.

ISM with 36 increments within the 175-ft grid cell would "hit" a 21-ft radius (42 ft diameter or 1/30<sup>th</sup> of an acre) 95% of the time.

![](_page_30_Figure_0.jpeg)

1 acre = 43,560 sq ft

![](_page_31_Picture_0.jpeg)

![](_page_32_Picture_0.jpeg)

- I'm going to cover two case studies in MO that used an ICS sampling design.
- Hopefully you will recognize some of the concepts Bob and Deana introduced in the first two modules.
- One is multi-county site in NW MO where tannery waste was used as a farm field fertilizer, and the other is a small historic lead smelter in south-central MO.

![](_page_33_Figure_0.jpeg)

- The red star on this map shows the location of the tannery in St. Joseph, MO along the MO River.
- The yellow polygons are locations of WWTP sludge application between 1983-2009
- 100+ farm fields, >10,000 acres affected
- Concern is about residual levels of hex Cr in sludge and whether they pose a health threat.

![](_page_34_Picture_0.jpeg)

- Sludge from facility WWTP was stored in bays prior to loading on trucks
- Transported to farms and spread at no cost to farmers

![](_page_35_Figure_0.jpeg)

- In pastures and no-till row crop fields, you can actually see the sludge residue
- This photo shows the visual fingerprint of sludge in a pasture

![](_page_36_Picture_0.jpeg)

- Here is our study question.
- EPA Region 7 developed a site-specific screening level for us of 86ppm
- Interested in the fine particulates since those most available via direct contact and most likely to be transported in dust.

Miss N	Missouri Department of Natural Resources																
	CONCEPTUAL SITE MODEL JACKPOT																
	Quarter	Section	Township	Range	Coord_N	Coord_W	County	Acres	Date Completed	Date Completed	Wet_Tons	Dry_Tons	Dry Tons Ac	Dry Tons Accum	Lbs Chr Acre	Crop Planted	
	NW	22	58N	32W	39 48.7	-94 37.4	Dekalb	175	2/25/200 5	3/9/2005	1584	443	2.53	2.53	0.67	Milo	
	NW	22	58N	32W	39 36.7	-94 41.4	Dekalb	175	3/1/2005	3/9/2005	1120	313	1.79	1.79	0.49	Milo	
	sw	20	59N	31W	39 48.7	-94 36.7	Dekalb	20	8/2/1999	8/5/1999	363	143	7.22	7.22	12.8	Grass/Hay	
	NE	30	59N	31W	39 36.2	-94 39.8	Dekalb	70	5/15/199 9	5/24/199 9	911	367	5.26	5.26	6.31	Grass/Hay	
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- You are not intended to read this table.
- It is only to illustrate the extensive data available from the tannery regarding sludge applications.
- This data aided us in developing a CSM and selecting fields to screen based on worst case conditions

![](_page_38_Figure_0.jpeg)

- We used the CSM to identify a subset of worst case farm fields to assess. If they came back clean, we would have high confidence that the others were OK as well.
- This is the same map as we saw before with yellow polygons being ffs that got sludge.
- The inset maps on this aerial photo show two individual farm fields. Each field was considered a DU
- Dus based on Eus ~ 80 acres
- · Goal is to obtain estimate of mean for the DU
- To make the sampling more manageable, the Dus were divided into 1-acre SUs. This is the scale at which ICS will be collected.

![](_page_39_Figure_0.jpeg)

- We conducted a DMA (pilot study) This is not something that would always be warranted, but it was in this instance due to the scope and complexity of this investigation.
- List the purposes and refer to vario-plot in photo.

![](_page_40_Figure_0.jpeg)

- The XRF would be useful DL was low enough to see Cr in fields
- · Our varioplots showed that total Cr and hex varied similarly across fields
- Problems with lab method identifed and corrected before full scale implementation
- · We got some initial ideas about sampling density

![](_page_41_Figure_0.jpeg)

- In the next few slides, I'm going to step you through the design
- Our ICS will be collected in the 1-acre SUs, so the first step is to determine how many increments in each ICS
- · We did this by collecting the increments as discrete samples
- DMA told us 10 would be in the ballpark so we collected that #
- Then we analyzed incr. total Cr by XRF and used VSP to see if 10 were enough
- If not, collect more incr.
- When we had enough, we combined all the incr. together to form the SU ICS
- This was repeated in 2 other Sus, so that we ended up with 3 bags of soil
- This is a little diff than the examples of composite averaging Deana showed Tuesday on Module 2. Here, we are not sampling in all portions of the DU. Instead, we are only sampling in some and then using the mean and SD to see

![](_page_42_Picture_0.jpeg)

This photo shows us flagging out a 1 acre SU prior to increment collection We used GIS tools in the office to lay 1 acre sampling grids across the Dus, loaded that onto GPS units and used those in the field to find and flag SU corners

![](_page_43_Figure_0.jpeg)

- This is a screen shot of VSP showing the results for a single 1-acre SU.
- Again, the VSP analysis was all done on total Cr results and was used only to determine sampling density.
- We entered in our level of confidence (95%), and then the WGR which is just the action level minus the measured average from the discrete samples.
- We need an action level for total Cr, but we only had one for hex Cr, so we had to convert hex Cr AL using 3x worst case ratio observed during DMA.
- We also enter in the mean and SD from the discretes and VSP provides an estimate of how many samples are needed to make this decision.
- Note in this case it is less than 10 so we can simply combine the 10 discretes already collected to form our ICS
- This was then repeated from the other two 1-acre Sus.

![](_page_44_Figure_0.jpeg)

- The second step in our design was to determine how many of the 1 acre SUs we needed to sample in the DU
- Our DMA told us 5 would be pretty close, so we started with the 3 bags of soil from step 1.
- As before we analyzed each SU ICS by XRF and calc. mean and SD.
- VSP was used to determine whether 3 SUs were enough or if we needed to go get more.
- Once we had enough, all SU ICS sent to lab for Cr 6 analysis
- For our final decision, a 95% UCL on the mean Cr6 conc. Was calculated from the SU ICS results.

![](_page_45_Figure_0.jpeg)

The photos on this slide illustrate our sample processing proceedure.

Samples were air dried and then disaggregated manually using a mallet

The samples were then sieved to .25mm to obtain the fine fraction particle size of interest prior to XRF & lab analysis

![](_page_46_Figure_0.jpeg)

- We sampled in 60 SUs located in 20 farm field Dus.
- No Sus needed more than 10 increments and no farm fields needed more than three 1-acre Sus
- The key factor to this low sampling density was the distance between the mean we measured and the SL. Recall that this is called the WGR in VSP.

![](_page_47_Figure_0.jpeg)

- This slide shows the calculation of 95% UCL for a particular farm field
- Lab results for Cr6 were used to calculate a mean and SD which were entered into this equation for UCL using the student's t
- For this DU, we calculated a mean of 3.3 and a UCL of 5.3 ppm (far below the 86 ppm SL)

![](_page_48_Figure_0.jpeg)

- In some of the Sus, we collected replicate ICS to measure overall sampling and analytical precision
- Three separate ICS consisting of 10 increments were collected as shown in this diagram
- A RSD was calcuated from the lab results of the replicates as shown in the far right column of this table.
- Conventionally, ICS replicates are collected across an entire DU and used to calc a UCL. Here though, we used it for QC as a measure of overall sampling and analytical error since we were using the SU ICS to calc. the UCL.
- It was at below the 30% criteria established for this project in all cases indicating that the overall sampling and analysis error were well controlled

Natural Resourc	es.									
	RY D	UPLI	CA	ΓES						
by lab represent the entire	jar	DU-SU	Cr⁺ <sup>6</sup>	Cr <sup>+6</sup> Lab repl.	RPD					
of soil?		201	0.06	0.063	4.9					
	_	209-96	1.05	0.942	10.8					
1 the same	1	205-88	1.110	0.938	16.8					
0.1g 0.1g	212-53	0.605	0.750	21.4						
0.10		213-44	0.833	0.788	5.6					
0.1g		214-25	1.610	1.620	0.6					
0.1g 0.1g		215-55	0.069	0.080	14.8					
0.10	6	217-103	0.793	0.775	2.3					
0.1g		221-162	4.38	3.76	15.2					
0.1g		L								
C	lu-In Increm	ental-Composit	e			_ /				
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- We were also interested in what portion of the overall sampling and analysis precision was coming from subsampling procedure at the lab.
- The lab needed 1 gm for digestion/analysis.
- Rather than the typical "dig a spot" method we requested they obtain the 1gm analyzed from 10 x 0.1g subsamples collected thruout the sample mass. This process is shown in the photo to the left.
- The table to the right shows the results from the lab replicate pairs and the %RSD between them.
- You can see from the low RSDs that the subsampling process contributes relatively little to the overall sampling error. So this was not turn an important source of error for this site, but it may be at sites where the overall S & A precision is interferring with your ability to make a decision, and you need drill down and find out which step in the process needs to be improved.

![](_page_50_Figure_0.jpeg)

The key points about this case study are

- 1. the use of a strong CSM to focus on worst case farm fields and then have confidence making decision for the others
- 2. Using a pilot study revealed problems with our analytical method which were corrected before implementing full scale sampling.
- 3. The use of field analytical (XRF) together with an ICS strategy streamlined our investigation saving time and money.

![](_page_51_Figure_0.jpeg)

The second case study is a small historic lead smelting operation located near what is now a campground in a public recreational area.

![](_page_52_Figure_0.jpeg)

Some historic records available

Operated for 2 years in late 1800s. Minor quantities of lead were produced. Approximate location of smelter known from an old sketch, but no visual evidence remains.

![](_page_53_Picture_0.jpeg)

Photo in lower left shows a shaft.

Photo in upper right shows a campsite near where we believe the smelter was located.

![](_page_54_Picture_0.jpeg)

Here is our study question.

This is investigation has not been completed yet. State risk assessors are currently developing site-specific screening levels for recreational use.

![](_page_55_Figure_0.jpeg)

- 30 incr. recommended default to capture spatial variability across a DU when no estimates are available
- 3 replicate ICS allows calc. of a SD which can be used to develop a UCL
- With our 9 Dus, this would require a great deal of field work
- Question the need this is a screening level assessment of a lightly used recreational site on remote public land

![](_page_56_Figure_0.jpeg)

- Chose to use in-situ XRF to get initial estimates of variability and ranges of concentrations
- Also can help identify areas of the campground that may have elevated conc. Such as those near the former smelter.
- Use that data to streamline ICS sample collection

![](_page_57_Figure_0.jpeg)

- This figure shows the results of in-situ XRF analyses conducted in the 6 campgrounds. There was an additional area not shown here that is planned for development as a future campsite.
- Also, some of these Xs represent locations where we shot shallow (2-12") depth soil. Lead conc. Trends of surf followed those of subsurf.

![](_page_58_Figure_0.jpeg)

- Based on the XRF results, we developed an ICS design in which we would collect fewer increments (15vs. 30) in the campsite Dus where we saw lower concentrations and variability. We also collected fewer incr. for the subsurface (2-4") depths, primarily due to practical constraints of digging holes.
- Additionally, we would collect 3 replicates in the two campsites where we suspect levels may be close to or exceed SLs, but none in the other campgrounds or in the depth Dus.
- These choices reduce the level of effort required in the field, but as you will see there are some consequences in higher data variability.

![](_page_59_Picture_0.jpeg)

- These photographs show surface and subsurface increment collection.
- We used one of the commercially available incremental sampling tools which allowed us to complete sampling in all DUs one day including in-situ XRF work.

![](_page_60_Picture_0.jpeg)

- This series of photographs show the sample processing steps.
- Immediately upon return to the lab, sample cores were disaggregated by hand.
- Middle photo shows before and after.
- They were then air dried for 2 days, returned to the sample bags and further disaggregated by pounding with a rubber mallet.
- Samples were then sieved to 0.25 mm and analyzed by XRF.

![](_page_61_Figure_0.jpeg)

- As Deana mentioned earlier, Pro UCL is not designed to work with ICS data and it will not calculate a UCL from only 3 data points. So, instead we used this standard equation for t-test UCL as we saw in the previous case study.
- First thing to note here is how much higher these concentrations are than what we saw with in-situ XRF which were all <600ppm. Two reasons: soil was wet, and bulk soil vs small particle size.
- Also, notice how close the ICS replicate results are to each other indicating excellent precision.
- Because of this low SD, the UCL is only about 8% higher than the mean.
- This campsite will likely require a cleanup based on SLs used at other rec. sites.

![](_page_62_Figure_0.jpeg)

- There is a problem with this using this approach for the subsurface samples in Dus 1&2. We didn't collect replicates.
- Without an estimate of est. of var, cannot calc a UCL.
- But if we assume that lead is distributed in the subsurface similarly as in the surface, we can derive an SD using the relationship of the between-reps SD and the between-incr. SD as shown in these two steps
- The equation below step 1 on this slide shows the relationship between the SD between estimates of the mean (repl.) and the SD between increments making up the ICS.
- We can use this to derive a SD for 3 simulated replicate subsurface ICS.
- Also we only collected 15 incr/DU vs. 30.
- In Step 2, we can rearrange this same equation to obtain an estimate of SD for a simulated 15-increment ICS.

![](_page_63_Figure_0.jpeg)

- Now that we have an estimate of variability for the subsurface Dus, we can use the same equation as before to calc a UCL.
- Notice that the UCL is about 20% higher than the estimate of the mean; about 20% vs. only about 8% higher in the surface ICS
- That is the penalty for not collecting subsurf replicates and having to settle for a very conservative estimate of variability from surf samples.

![](_page_64_Figure_0.jpeg)

- Now, looking at the other campsites, again, we have no replicates for Dus 3-6 so no way to calc a UCL.
- Based on what we saw with the in-situ XRF data, it is less likely that the conc. And variability in Dus 1&2 are similar to those in 3-6, so not as defensible to extrapolate SD from 1&2.
- So instead, we can treat the entire 3-6 group as a separate population and use Pro UCL tools to develop a UCL for the group.
- The group UCL can then be used to derive uncertainty estimates for each individual DU.
- So let's step through that.

![](_page_65_Figure_0.jpeg)

- First it is important to know what sort of distribution we might be dealing with.
- ProUCL has a tool that will generate Q-Q plots.
- A QQ plot is a just a graphical way to look at your data. It plots the data against quantiles generated from a theoretical normally distributed population.
- A straight 45 deg line indicates the data is from a normal distribution.
- A break in the slope can indicate non-normal distr. Or that there are two separate populations.
- The graph to the left is the raw data for each DU ICS. Not normal.
- The graph to the right is for log-transformed data. Straight line here suggests the data are lognormally distributed
- ProUCL can calculate UCL for a number of different distributions incl.
   lognormal

![](_page_66_Figure_0.jpeg)

- This slide is a screen shot of Pro UCL output.
- A UCL for the group of DUS 3-6 is calculated at about 500ppm
- Notice that is about 50% higher than the mean for the group.

![](_page_67_Figure_0.jpeg)

- Now we can pull this all together.
- We can use the difference between this group UCL and group mean as a measure of uncertainty for the data set.
- We can then add that measure of uncertainty to each DU ICS result to obtain an conservative estimate of the mean for each DU as shown in the rightmost column of this table.
- Notice how much higher these estimates are than the individual ICS results. Up to 70% higher for DU 4.
- So there is a big penalty paid in data variability here for not having collected replicate ICS in each DU.
- However, if the SL comes out above these conservative estimates of the mean, this penalty will not interfere with our decision to conclude the campsites do not pose a risk.

![](_page_68_Figure_0.jpeg)

The key concepts for this case study are

•The use of FAM can refine a CSM and direct sampling efforts

•Using fewer increments and replicates saved time in the field and lab which can be a big factor with limited budgets, but

•Penalty to be paid in higher estimates of the mean

![](_page_69_Figure_0.jpeg)