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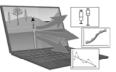
- Technical and Regulatory Web-based Guidance on Groundwater Statistics and Monitoring Compliance (GSMC-1, 2013) at <u>http://www.itrcweb.org/gsmc-1/</u>
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2



## Groundwater Statistics for Environmental Project Managers



Groundwater Statistics and Monitoring Compliance (GSMC) Technical and Regulatory Guidance Web-Based Document (GSMC-1)

Sponsored by: Interstate Technology and Regulatory Council (<u>www.itrcweb.org</u>) Hosted by: US EPA Clean Up Information Network (<u>www.cluin.org</u>)

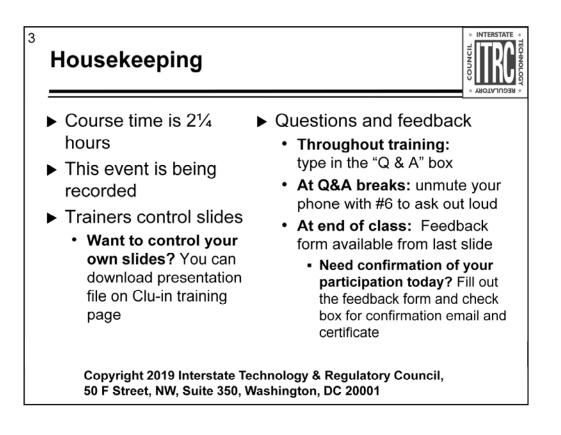
Statistical techniques may be used throughout the process of cleaning up contaminated groundwater. It is challenging for practitioners, who are not experts in statistics to interpret and use statistical techniques. ITRC developed the Technical and Regulatory Web-based Guidance Document on Groundwater Statistics and Monitoring Compliance (GMSC-1, 2013) and this associated training specifically for environmental project managers who review or use statistical calculations for reports, who make recommendations or decisions based on statistics, or who need to demonstrate compliance for groundwater projects. The training class will encourage and support project managers and others who are not statisticians to: -- Use the ITRC Technical and Regulatory Web-based Guidance Document on Groundwater Statistics and Monitoring Compliance (GMSC-1, 2013) to make better decisions for projects -- Apply key aspects of the statistical approach to groundwater data -- Answer common questions on background, compliance, trend analysis, and monitoring

-- Answer common questions on background, compliance, trend analysis, and monitoring optimization

ITRC's Technical and Regulatory Web-based Guidance Document on Groundwater Statistics and Monitoring Compliance (GMSC-1, 2013) and this associated training bring clarity to the planning, implementation, and communication of groundwater statistical methods and should lead to greater confidence and transparency in the use of groundwater statistics for site management.

ITRC (Interstate Technology and Regulatory Council) <u>www.itrcweb.org</u> Training Co-Sponsored by: US EPA Technology Innovation and Field Services Division (TIFSD) (<u>www.clu-in.org</u>)

ITRC Training Program: training@itrcweb.org; Phone: 402-201-2419

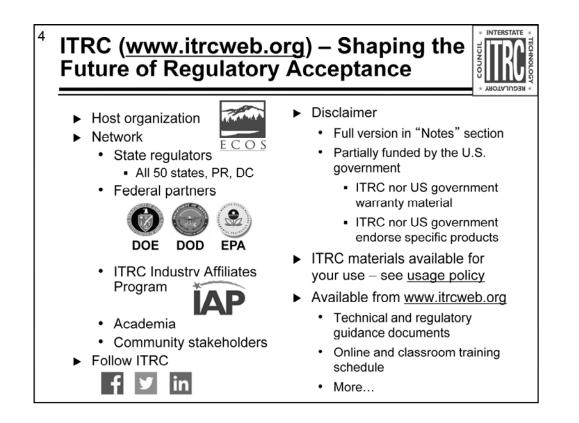


Although I'm sure that some of you are familiar with these rules from previous CLU-IN events, let's run through them quickly for our new participants.

We have started the seminar with all phone lines muted to prevent background noise. Please keep your phone lines muted during the seminar to minimize disruption and background noise. During the question and answer break, press #6 to unmute your lines to ask a question (note: \*6 to mute again). Also, please do NOT put this call on hold as this may bring unwanted background music over the lines and interrupt the seminar.

Use the "Q&A" box to ask questions, make comments, or report technical problems any time. For questions and comments provided out loud, please hold until the designated Q&A breaks.

*Everyone* – please complete the feedback form before you leave the training website. Link to feedback form is available on last slide.



The Interstate Technology and Regulatory Council (ITRC) is a state-led coalition of regulators, industry experts, citizen stakeholders, academia and federal partners that work to achieve regulatory acceptance of environmental technologies and innovative approaches. ITRC consists of all 50 states (and Puerto Rico and the District of Columbia) that work to break down barriers and reduce compliance costs, making it easier to use new technologies and helping states maximize resources. ITRC brings together a diverse mix of environmental experts and stakeholders from both the public and private sectors to broaden and deepen technical knowledge and advance the regulatory acceptance of environmental technologies. Together, we're building the environmental community's ability to expedite quality decision making while protecting human health and the environment. With our network of organizations and individuals throughout the environmental community, ITRC is a unique catalyst for dialogue between regulators and the regulated community.

For a state to be a member of ITRC their environmental agency must designate a State Point of Contact. To find out who your State POC is check out the "contacts" section at www.itrcweb.org. Also, click on "membership" to learn how you can become a member of an ITRC Technical Team.

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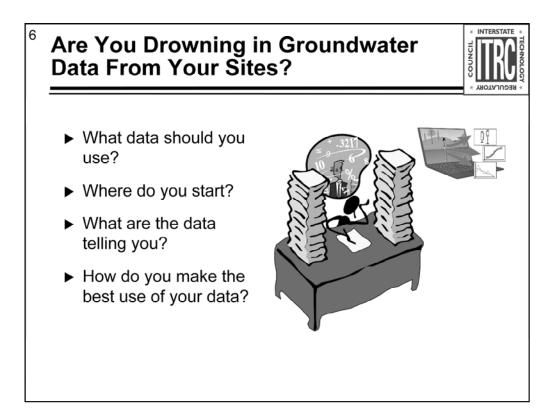


Harold Templin is a Senior Geologist at the Office of Land Quality of the Indiana Department of Environmental Management (IDEM). Since 1985, Harold has worked as a geologist with compliance and permitting of hazardous waste facilities and solid waste landfills. In that role, Harold has designed or approved ground water monitoring systems, sampling and analysis plans, and statistical evaluation plans. While working at IDEM Harold has set on numerous panels setting policy and guidance for closure of land disposal facilities, and risk assessment of corrective action sites. Since 1977, Harold has worked with land use regulations and with efforts to encourage the best use of our natural resources. This is Harold's first effort of working on an ITRC team. Harold earned a BS degree in geology from Indiana State University in 1972. Additionally, Harold has taken additional graduate level courses in hydrogeology and groundwater modeling at Indiana University. Harold is a licensed professional geologist in the State of Indiana.

**Chris Stubbs** is a principal consultant with Ramboll in Emeryville, California. Since 2000, he has worked in environmental science and engineering, with special emphasis on groundwater hydrology and chemical fate and transport in the environment. Specific areas of expertise include groundwater modeling, statistical analysis, risk-based site assessment and remediation, exposure analysis and human health risk assessment. He has prepared evaluations of the risk from vapor intrusion into indoor air at numerous sites, including preparing expert reports and giving deposition testimony as an expert witness. He has developed regional groundwater flow and transport models to evaluate remedial alternatives and to estimate cleanup times. Chris is a member of the ITRC Groundwater Statistics and Monitoring Compliance project team. Chris earned a bachelor's degree in 1988 in physics from the University of California at Berkeley, CA. He earned a master's degree in 1996 in technology and policy, a master's degree in 1996 in environmental engineering, and a PhD in 2000 in hydrology and water resource engineering all from the Massachusetts Institute of Technology in Cambridge, MA. He is a professional civil engineer in California.

Liz Simmons is a Senior Principal Geologist and Senior Fellow over the Environmental Site Characterization Practice for Kleinfelder in the San Diego, California office. Liz has worked for Kleinfelder since 1998, and in the environmental field since 1984. Her extensive technical background includes numerous groundwater assessments, risk assessments, modeling, groundwater optimization, and litigation support for sites encompassing industrial facilities, landfills and surface impoundments, oil refinery and underground fuel storage sites, and superfund sites, both in the United States and abroad. As Technical Practice Leader, she is also responsible for providing on-going technical training to Kleinfelder personnel. Liz has been active in the ITRC since 2008, and has been involved with the Contaminated Sediments, Groundwater Statistics and Monitoring Compliance, and Geophysical Classification for Munitions Response workgroups. She earned a bachelor's degree in geology from Brigham Young University in Provo, Utah in 1979, and underwent graduate studies in petroleum geology and geophysics at California State University in Long Beach from 1981-1984. Liz is a registered PG with the California Board for Professional Engineers, Land Surveyors and Geologists.

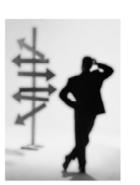
**Randall Ryti** is a Senior Scientist with Neptune and Company, Inc. in Los Alamos, New Mexico. Randall was one of the founding partners of Neptune in 1992 and has worked on basic and applied environmental problems, including ecological and human health risk assessment support to the Los Alamos National Laboratory, the Hanford Site, and other clients. These risk assessment projects have addressed environmental contamination including groundwater over large spatial scales with multiple stressors, large and complex databases, and significant involvement of regulators, natural resource trustees, and the public. Randall has also assisted in the planning, decision logic, and statistical design of environmental data collection activities for Environmental Restoration sites at Department of Energy facilities (Hanford, Los Alamos, and Savannah River), Department of Navy Base Realignment and Closure facilities in California, and EPA-lead Superfund sites. Randall has been a member of ITRC teams since 2009 and has been active in the Groundwater Statistics and Monitoring Compliance team starting in 2011. Randall received the Industry Recognition Award from the ITRC in October 2012. Randall earned a bachelor's degree in biology from University of California Los Angeles in 1979 and a PhD in biology from University of California San Diego in 1986. Randall is certified as a Senior Ecologist by the Ecological Society of America.



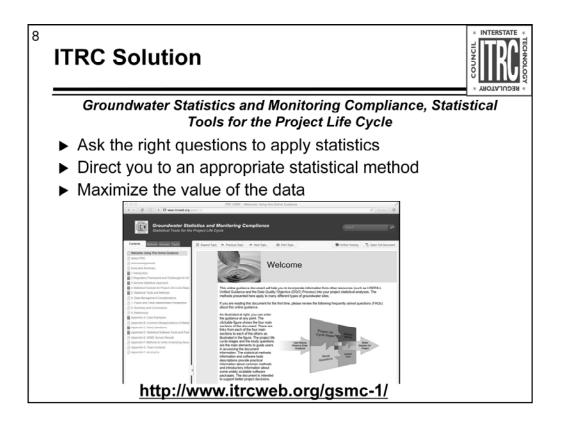


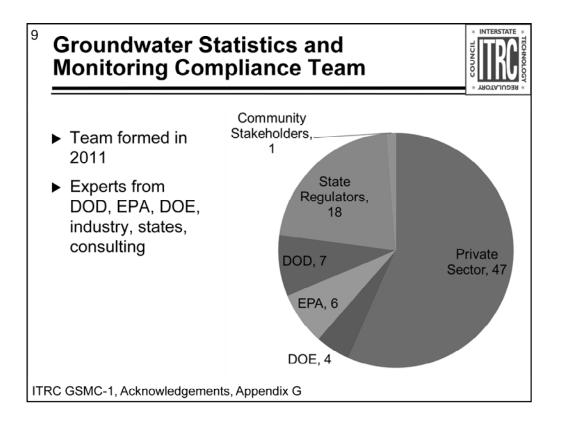


- ▶ If you are not a statistician
  - · More informed consumer of statistics
  - Confidence to spot misapplications and mistakes
  - Review selection of tests
  - Understand language of statistics
- If you are a statistician
  - Help make your work and conclusions understandable to a general audience



ITRC GSMC, Section 1

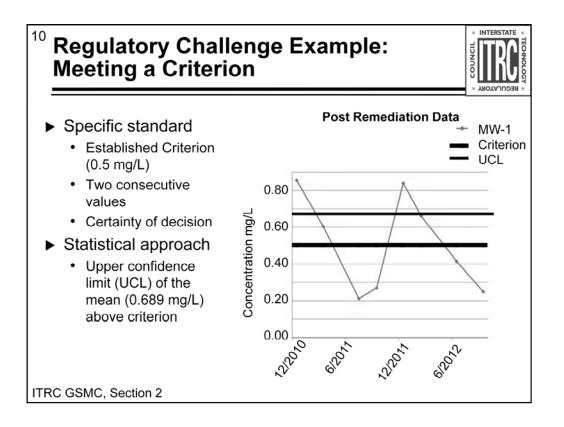




ITRC formed this team in 2011 to develop the ITRC Groundwater Statistics and Monitoring Compliance document and training.

The team of experts includes

- 1. State regulators
- 2. Federal partner experts from DOD, DOE and EPA
- 3. Consulting community and from industry

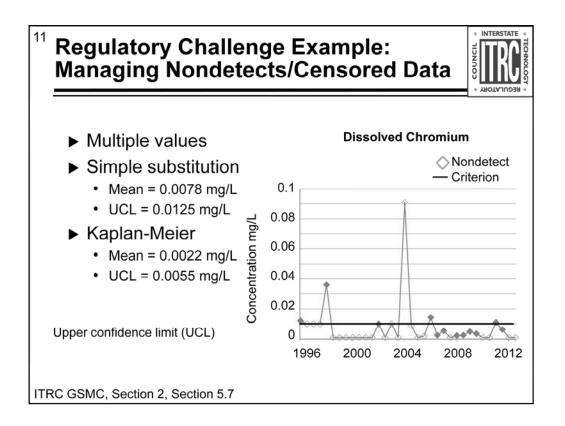


Statistically, the upper confidence limit (UCL) is 0.689 mg/L more remediation is needed to be confident the groundwater concentration is below the criterion.

Additional reference for information about groundwater variability: McHugh, T., L. M. Beckley, et al. (2011). "Factors influencing variability in groundwater monitoring data sets." Ground Water Monitoring & Remediation 31(2): 92-101.

Definition of upper confidence limit (UCL): The upper value on a range of values around the statistic (for example, mean) where the population statistic (for example, mean) is expected to be located with a given level of certainty, such as 95% (science-dictionary.org 2013).

See the GSMC-1 document for more glossary items: www.itrcweb.org/gsmc-1



Definition of nondetects: Laboratory analytical result known only to be below the method detection limit (MDL), or reporting limit (RL); see "censored data" (Unified Guidance).

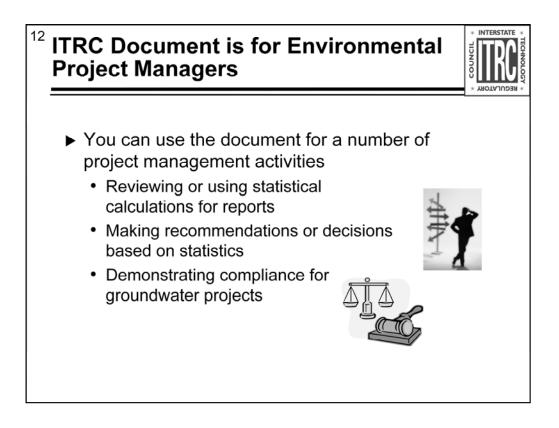
Definition of censored data: Values that are reported as nondetect. Values known only to be below a threshold value such as the method detection limit or analytical reporting limit (Helsel, D.R. 2012)

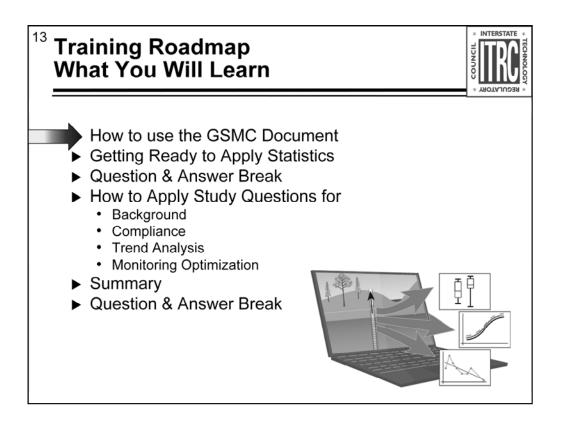
Helsel, D.R. 2005. Nondetects and Data Analysis. Hoboken, NJ: John Wiley & Sons.

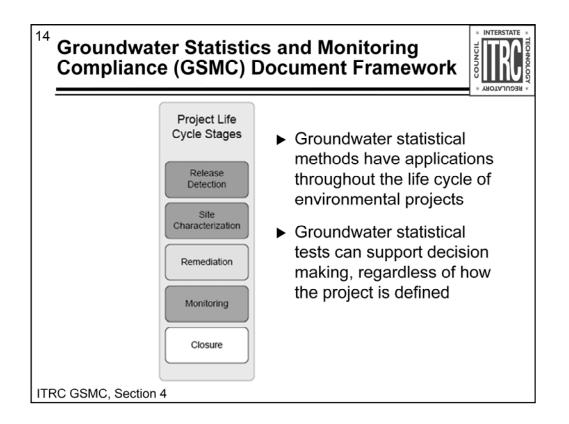
Helsel D.R. 2012. Statistics for Censored Environmental Data Using Minitab and R. 2nd ed. Hoboken, New Jersey: John Wiley & Sons.

USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency.

http://www.epa.gov/osw/hazard/correctiveaction/resources/guidance/sitechar/gwstats/unified-guid.pdf.

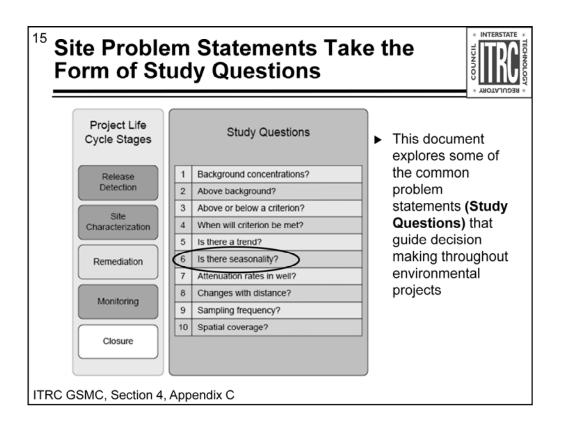


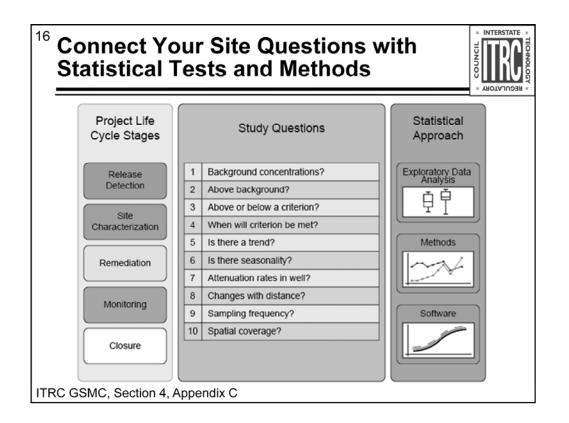




The terminology and framework for the project life cycle stages may vary under different regulatory programs. To simplify the organization of the document the team chose these 5 project life cycle stages.

Groundwater Statistics and Monitoring Compliance (GSMC) Document (GSMC-1)





The study questions serve as a bridge to connect life cycle-based activities on your sites with relevant statistical approach (exploratory data analysis, tests, and software) that may assist you with site decisions using groundwater data.

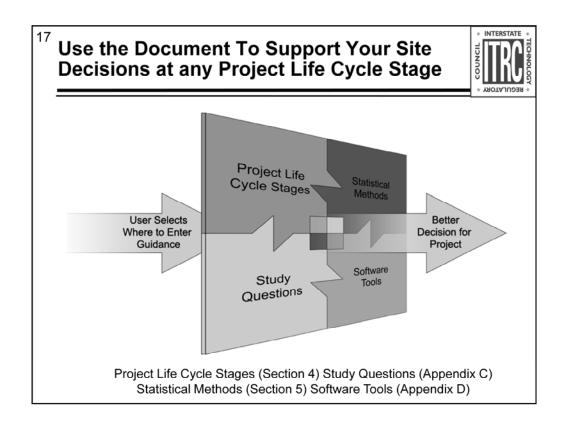
Definition of exploratory data analysis (EDA): An approach for initial data evaluation using graphical methods to open-mindedly explore the underlying structure and model of a dataset to aid in selection of the best statistical methods. Typical techniques are box plots, time series plots, histograms, and scatter plots (Tukey 1977; NIST/SEMATECH 2012; Unified Guidance).

NIST/SEMATECH. 2012. "e-Handbook of Statistical Methods." In. http://www.itl.nist.gov/div898/handbook/ (accessed August 2013).

Tukey, J.W. 1977. Exploratory Data Analysis. Reading, MA: Addison-Wesley.

USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency.

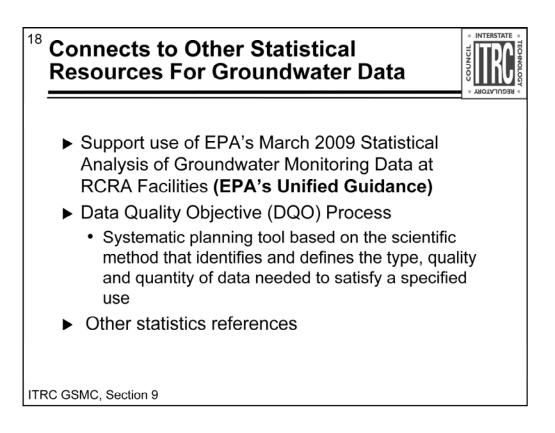
http://www.epa.gov/osw/hazard/correctiveaction/resources/guidance/sitechar/gwstats/unified -guid.pdf.



The web-based document was designed to be accessed from many different perspectives or sections

All of the sections of the document are interconnected and can be accessed from any other section.

This figure is located on the Welcome screen of the web-based tech-reg document and by clicking on the named parts of the document, you will go to that section of the document.



The team has tried to make this document user friendly for the Project Manager with its cross references and links to the study questions methods, and tools.

This web-based guidance document will help you to incorporate information from other resources

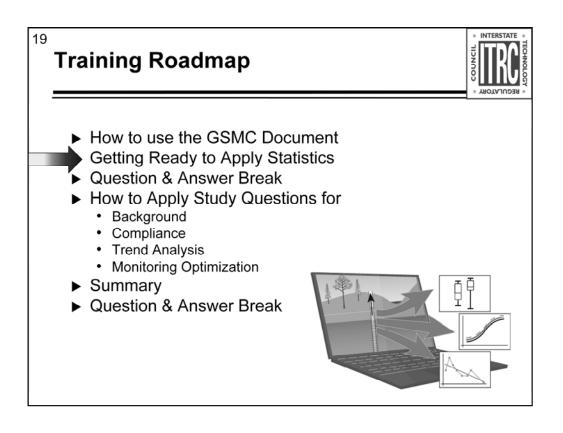
1.Unified Guidance (USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." In Unified Guidance EPA 530/R-09-007. Washington DC: Unites States Environmental Protection Agency. www.epa.gov)

2.Guidance on Systematic Planning Using the Data Quality Objective Process (USEPA. 2006a. "Guidance on Systematic Planning Using the Data Quality Objectives Process." In EPA QA/G-4. Washington D.C.: United States Environmental Protection Agency. http://www.epa.gov/QUALITY/qs-docs/g4-final.pdf)

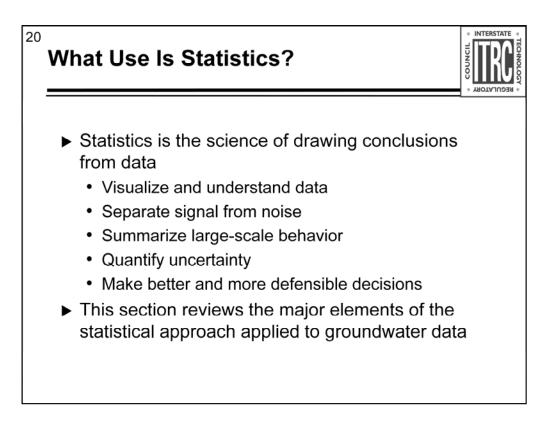
3.ASTM documents (for example, ASTM. 2012. Standard Guide for Developing Appropriate Statistical Approaches for Groundwater Detection Monitoring Programs. West Conshohocken, PA: ASTM International)

4.Textbooks on groundwater statistics (for example, Gibbons, R.D. 1994. *Statistical Methods for Groundwater Monitoring*. New York: John Wiley & Sons)

5.Other guidance documents on the DQO process (for example, United States Army Corp of Engineers (USACE). 1998. "Environmental Quality, Technical Project Planning (TPP) Process." EM 200-1-2. Washington, D.C.: Department of the Army. www.usace.army.mil/missions/environmental/technicalprojectplanning.aspx.)



This next section reviews the major elements of the statistical approach applied to groundwater data.



## <sup>21</sup> Hypothesis Testing: Guilty or Not Guilty?



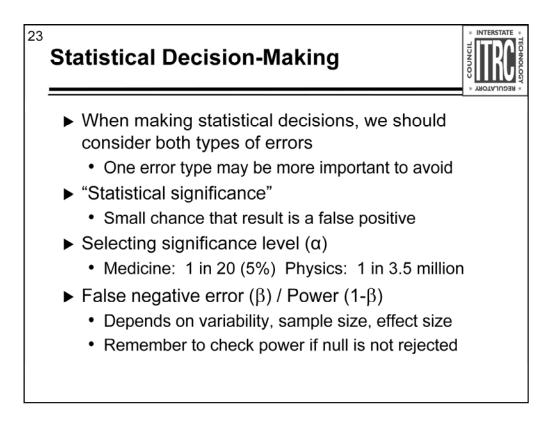
► Crime: Burglary (laptop)

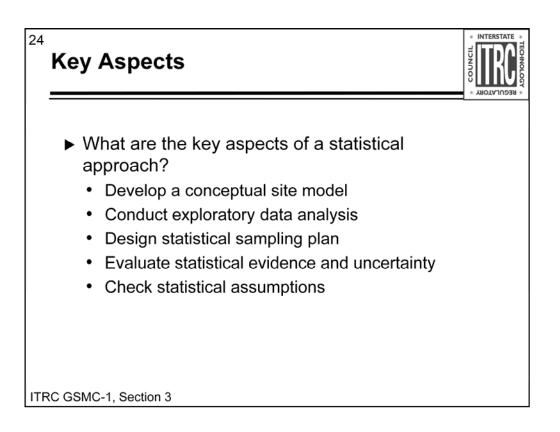


- ► Defendant: Mr. Ivan M. Shifty
- Assume defendant is not guilty (<u>null hypothesis</u>)
- Present evidence (fingerprints, stolen goods)
- Prove guilt beyond a reasonable doubt?
- We want to avoid errors
  - 1. Shifty is innocent but goes to jail (false positive)
  - 2. Shifty is guilty but goes free (false negative)

22	Decision Errors			
	Site characterization phase of life cycle Null hypothesis H <sub>0</sub> : Site groundwater is NOT contaminated			
	Decision based on statistical sample			
	True State of Site	H <sub>0</sub> : Site Not Contaminated	H <sub>A</sub> : Site Is Contamin	ated
	Not Contaminated	<b>Correct Conclusion</b> (Probability = 1-α)	False Positive (Probability = α) Significance Level	
	Contaminated	<b>False Negative</b> (Probability = β)	Correct Conclus (Probability = 1-) <u>Power</u>	
ITRC GSMC-1, Section 3.6.1				

See also USEPA. 2006a. "Guidance on Systematic Planning Using the Data Quality Objectives Process." In EPA QA/G-4. Washington D.C.: United States Environmental Protection Agency. http://www.epa.gov/QUALITY/qs-docs/g4-final.pdf. for more information (Table 7)





All of these concepts are discussed in Section 3 of the web-based guidance document www.itrcweb.org/gsmc-1

Purpose of using a statistical approach is to make better, more defensible decisions (for example, has the site impacted groundwater?).

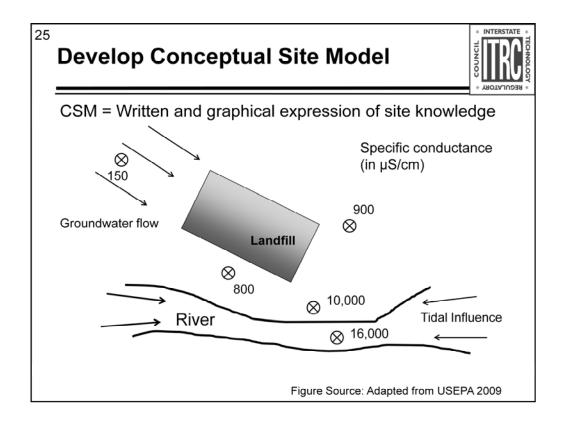
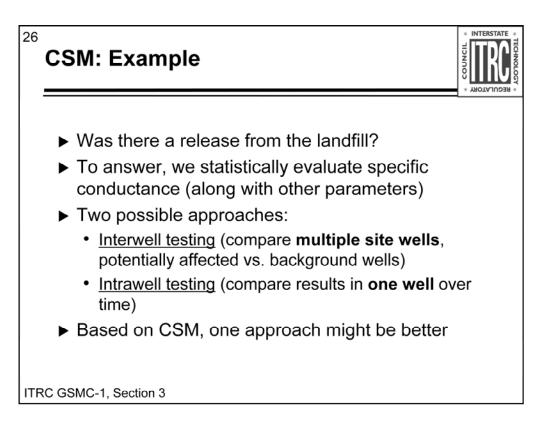
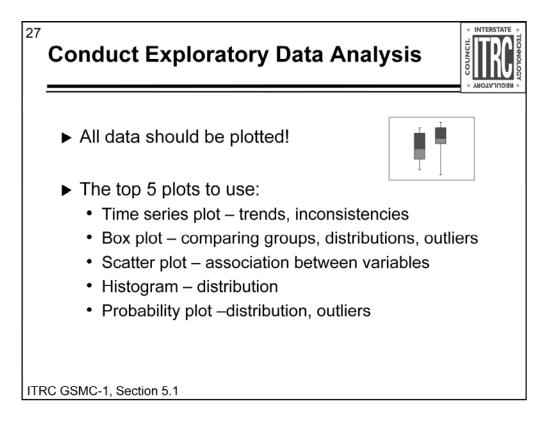
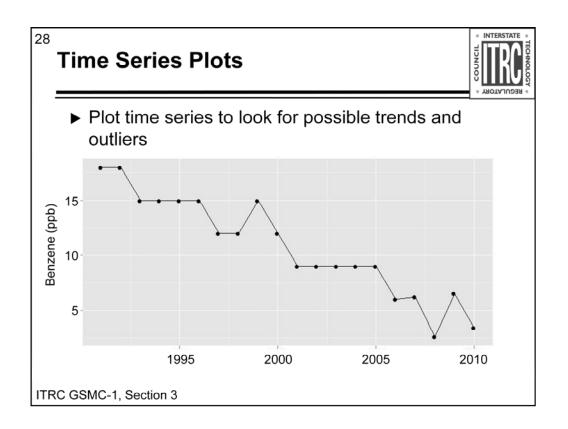


Figure Source: USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." In Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency. www.epa.gov

A CSM summarizes the potential contaminant sources and transport pathways through the subsurface. For example, here is a landfill situated near a stream. Groundwater flows under the landfill toward the stream. We want to look at specific conductance to see whether the landfill is leaking. But in this case, there is another potential source of chloride to groundwater—the stream is tidally influenced. Importance of getting the CSM right is crucial to doing good statistics.







Another important aspect of conducting a good statistical evaluation is exploratory data analysis (EDA). This is nothing more than taking a good hard look at the data, particularly using graphs that show clearly what's going on.

For example, we often want to look at trends. According to this plot, benzene concentrations at a well appear to be going down over time.

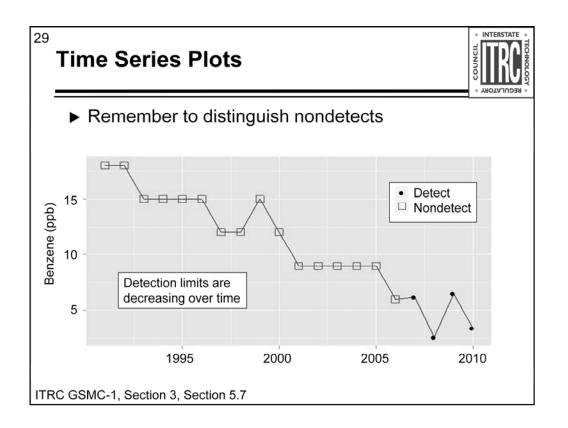
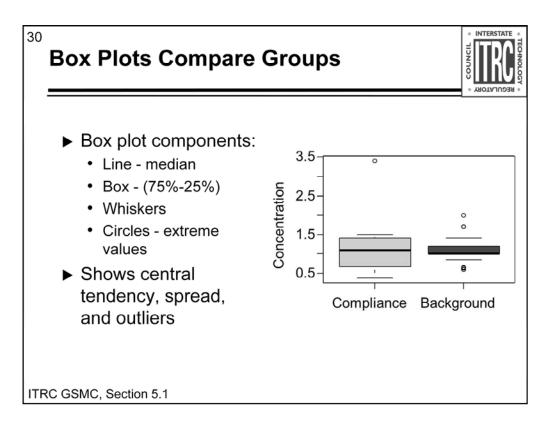


Figure 5-14 Example time series plot of benzene data with nondetects.



Understand your data; see what your data set is telling you. The side-by-side box plots show that the data sets have similar buy not equal variances. It is fairly common that data from a compliance well will have a larger variance than data from a background well.

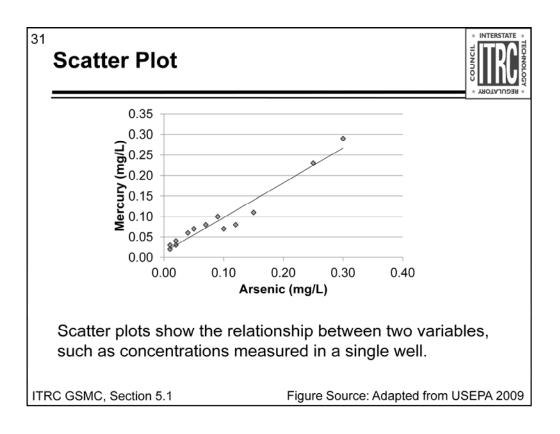
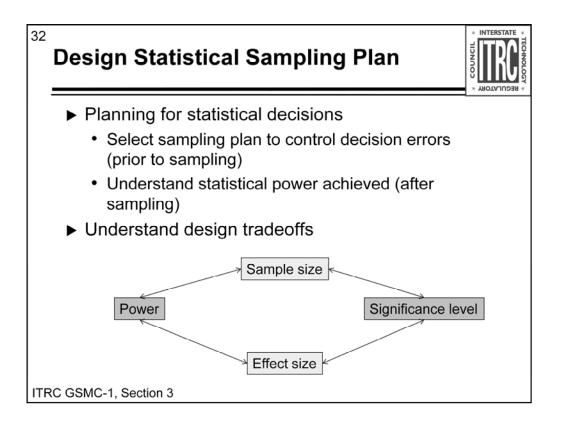
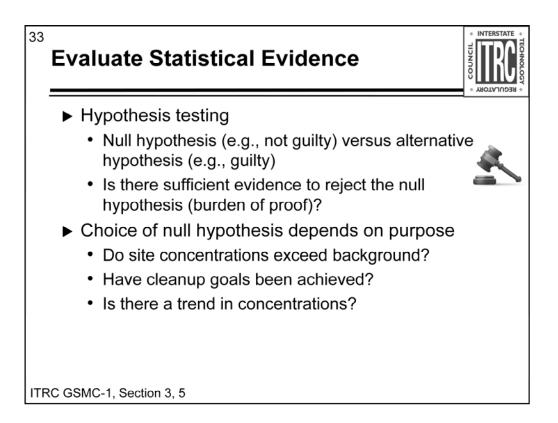


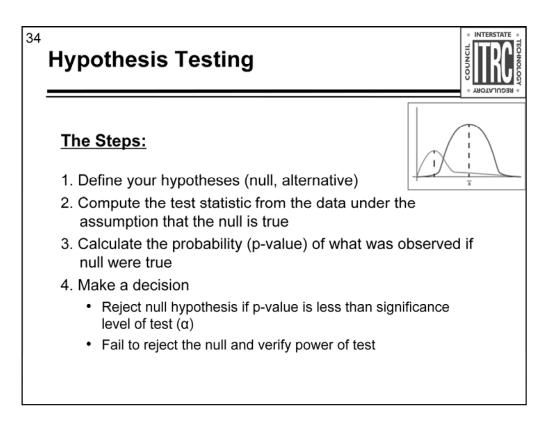
Figure Source: USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." In Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency. www.epa.gov

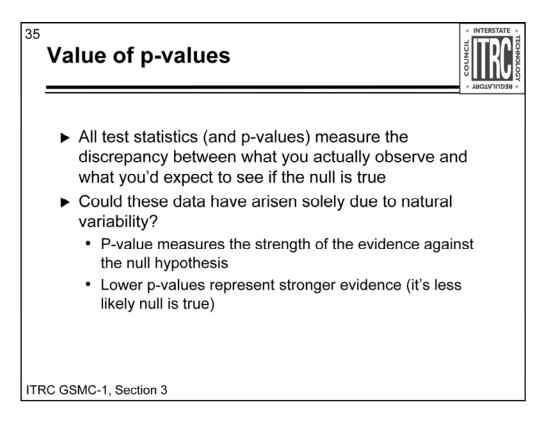


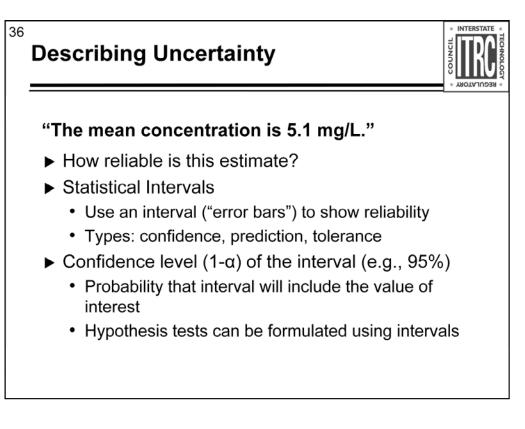
Every decision using real data involves uncertainty. Did we conclude the site was impacting groundwater when it really wasn't (false positive)? Or perhaps the site was a source of groundwater contamination and we didn't see it because there wasn't enough data (false negative). The best way to control these potential errors is through up front planning through a quality assurance project plan or something similar.

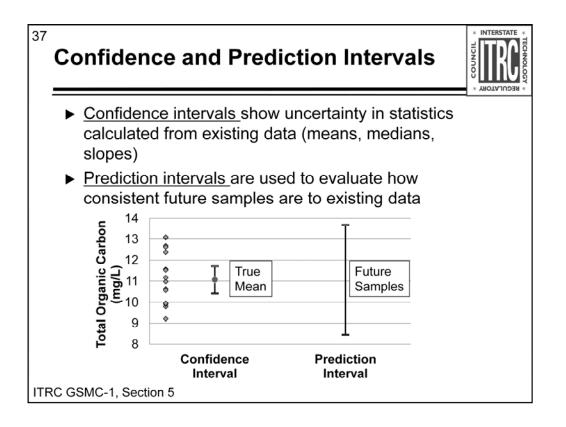
Sometimes we can't plan ahead. We can still use the same statistical tools to help us understand the level of uncertainty.











Confidence intervals are often used when comparing mean groundwater concentrations to fixed standards such as maximum contaminant levels (MCLs). Prediction intervals are often used for detection monitoring at RCRA facilities to determine whether future samples are consistent with background concentrations.

## <sup>38</sup> How to Select and Test the Null Hypothesis

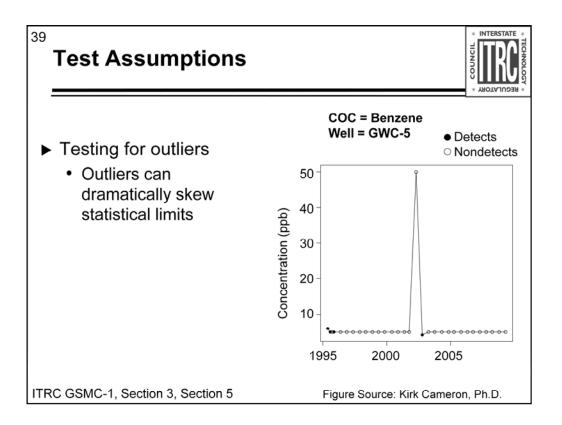


## Release Detection/ Site Characterization

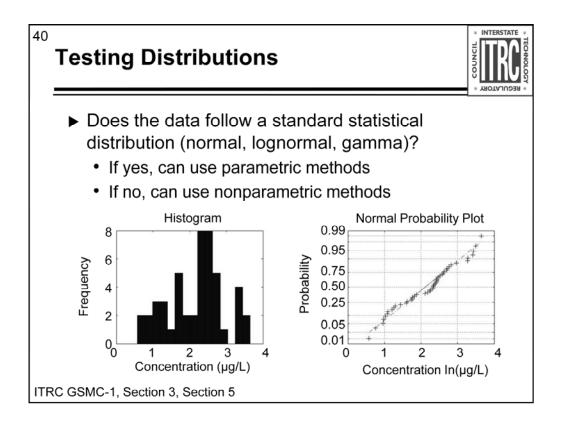
- Site is assumed to be clean unless proven otherwise
- ► Hypothesis:
  - Null: Mean ≤ MCL
  - Alt: Mean > MCL
- ► Confidence interval test:
  - Reject if <u>lower</u> confidence limit on mean > MCL

### Corrective Action/ Remediation

- Site is assumed to be contaminated unless proven otherwise
- ► Hypothesis:
  - Null: Mean ≥ MCL
  - Alt: Mean < MCL
- ► Confidence interval test
  - Reject if <u>upper</u> confidence limit on mean < MCL

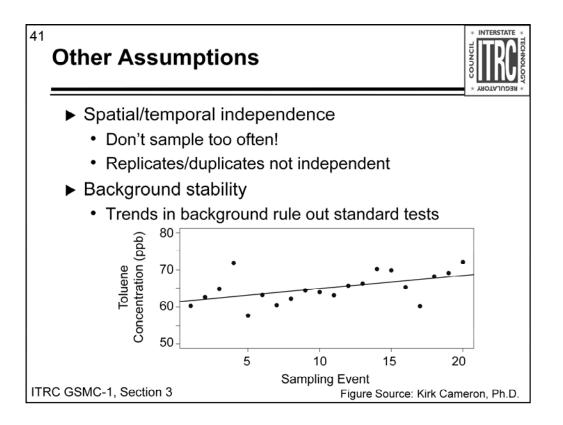


All statistical tests are based on some sort of simple model of reality. The statistical model never matches reality exactly, but we need to check that the assumptions of the model are more or less satisfied. For example, the presence of outliers can wreck havoc with statistical tests. These extreme values should be identified and investigated – just a typo, lab error, or maybe real.



Another common assumption is that of normality. Do the data follow a bell curve? If not, then non-parametric approaches should be favored.

Figure A.5 from the web-based document.



Temporal independence– standard tests assume independence; positively correlated data tend to underestimate true variance; can lead to wrong decisions

Advice- don't sample too often; quarterly sampling just a rule of thumb; run a pilot study to get site-specific sampling frequency

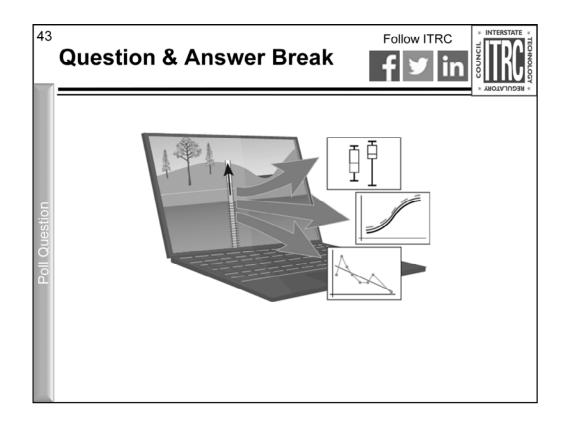
Background stability/stationarity- standard tests assume a stationary mean in background; variance will be overestimated and significant changes missed if trends are not accounted for

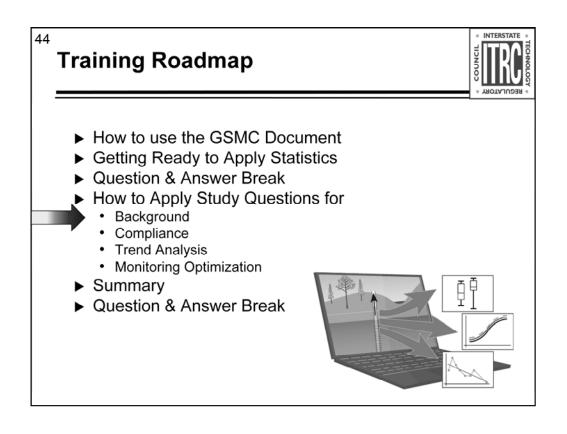
Advice- switch to trend test or remove effect of trend first

## <sup>42</sup> Closing Thoughts on Getting Ready to Apply Statistics



- Plot the data
- Are there sufficient data to make a good decision with appropriate error rates?
- ► Confirm regulatory requirements
- Use results of statistical analyses with other lines of evidence
- "If you torture the data long enough, it will confess to anything."

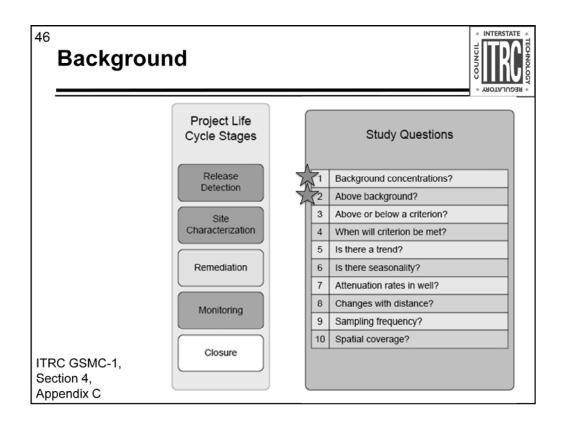




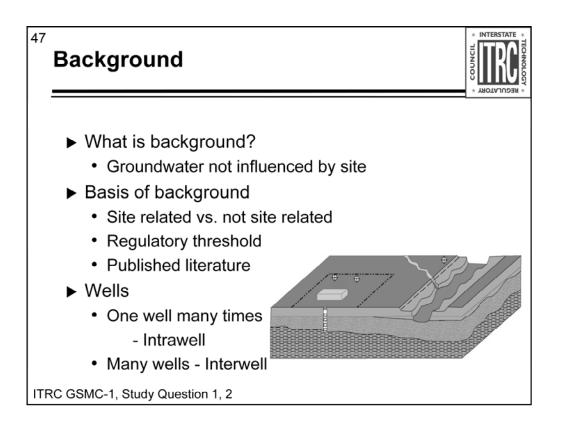
<sup>45</sup> Connect Study Q	ing Li uestio	fe ons	Су	cle	e S	tag	
	Proj	ect Life	e Cycle	e Stage	es		
	Release	Site Characterization	Remediation	Monitoring	Closure		Study Questions
	×	х			х	1	Background concentrations?
	x	х		Х	х	2	Above background?
	x	х	х	х	х	3	Above or below a criterion?
			Х	Х	х	4	When will criterion be met?
			х	х	х	5	Is there a trend?
			Х	Х	х	6	Is there seasonality?
			х	х	х	7	Attenuation rates in wells?
			Х	Х	х	8	Changes with distance?
			х	х	х	9	Sampling frequency?
TRC GSMC-1, Section 4,	x	х	х	х	х	10	Spatial coverage?

We use the project life cycle stages and the study questions to illustrate the connections to the example information for the hypothetical sites.

Project life cycle stages are shown as columns. The study questions are listed on the right. The study questions that generally apply to each of the life cycle stages are noted with the X's in the table on the left. In the web-based document this connection between life cycle stages and study questions is used to help the reader find information that is relevant for their particular site.



Two of the 10 questions are directly related to background.



Site related vs. not site related: In general, both native and man-made sources need to be taken into account; however, a baseline must be established, because groundwater may already be tainted with concentrations of a contaminant from either native or man-made sources that would make it difficult, if not impossible, to clean up to a "pristine" standard. The web-based guidance document addresses this in more detail.

Intrawell vs. Interwell (see Slide 31)

Intrawell – Sample one well many times and compare Interwell – Sample many wells many times and compare

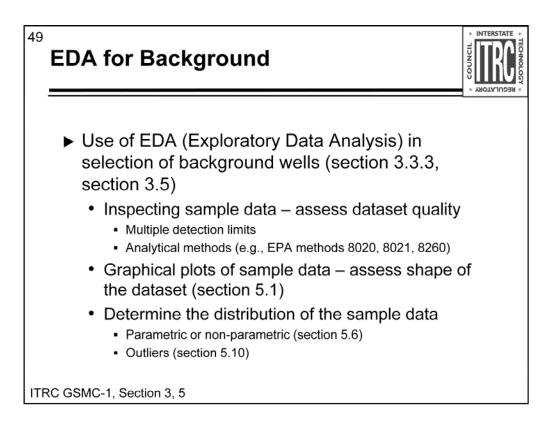
## <sup>48</sup> Considerations for Statistical Analysis of Background



- Avoid known sources of site-related contamination
- Distance and direction of selected wells from source
- Review of geologic/hydrogeologic information
- ► Multiple aquifer characteristics
- Project data quality objectives (DQOs)
  - Sufficient number of samples
  - Quality of the dataset

DQOs – should be developed up front Sufficient samples to provide reliable results Quality of dataset (multiple detection limits)

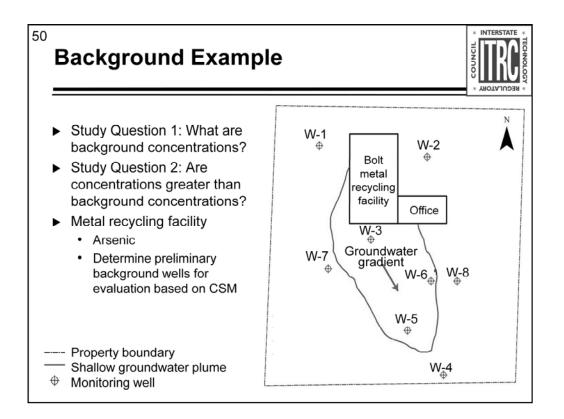
A good CSM model should be developed to maximize the validity of your assumptions.



Use Exploratory Data Analysis methods to help evaluate a potential background dataset.

To get a quick visual, use graphical methods such as histograms, quartile plots, box plots or scatter plots to plot the dataset and evaluate the shape of the dataset to provide insight into the overall spread and potential distribution of the data.

Distribution of the dataset will help you determine whether parametric or non-parametric (doesn't fit normal distribution) statistical methods will be needed in your evaluation of background. For example, t-test (parametric) is sensitive to outliers.



This example site shows a site with a groundwater plume and wells located upgradient (W-1 and W-2), side-gradient (W-7 and W-8) and down-gradient (W-3, W-4, W-5 and W-6) from the plume. It will be used to show how potential background well data can be evaluated to determine whether the dataset can be used to characterize background.

Background Examp	le Datase	t	* INTERSTATE
	Arsenic (µg	/I)	
	Date	W-1	W-2
Identify a dataset	1/2009	2.9	3.1
-	4/2009	3.1	4.9
Number of samples	7/2009	2.6	2.6
<ul> <li>Address nondetects</li> </ul>	10/2009	2.4	2.5
	1/2010	2.7	3.2
	4/2010	3.0	7.5
	7/2010	2.6	2.8
	10/2010	2.5	2.8
	1/2011	2.7	3.5
	EPA Drinking	g Water MC	L = 10 µg/l

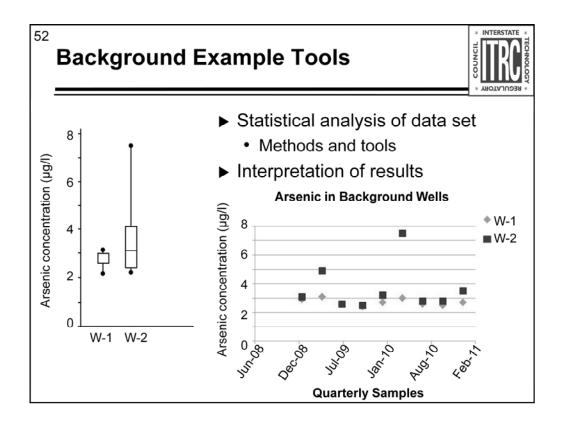
To identify whether a dataset that would be consistent with background:

Select a potential dataset - evaluate quality, shape and distribution.

Outliers?

What are the DQOs - how do they apply to number of samples? Are there non-detects (in this case, no)?

Both upgradient wells are below the MCL.



Methods and tools to do preliminary statistical analysis:

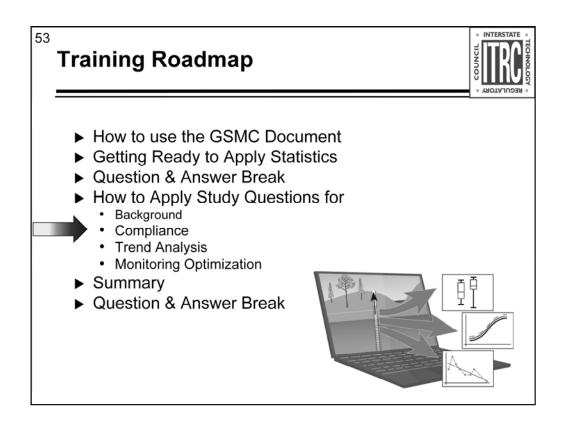
Probability Plots (Section 5.1.5), Time Series Plots (Section 5.1.1), Outlier Identification (Section 5.10)

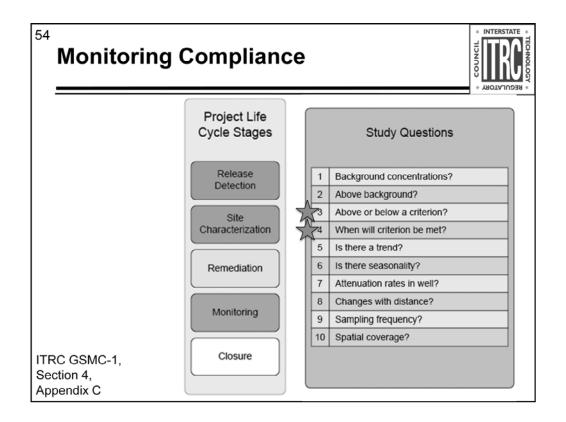
Interpretation of Results:

Understanding the source of variation (natural or man made, site related or not site related)

Temporal changes Presence of outliers

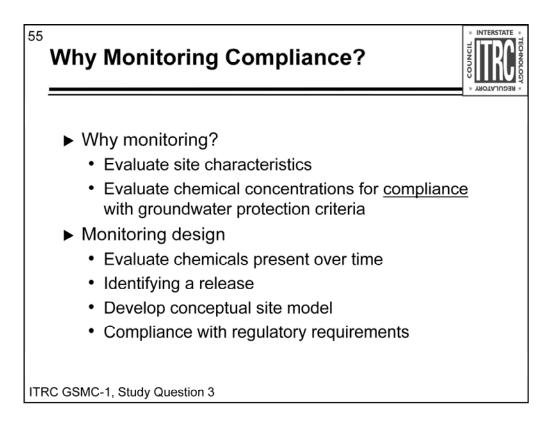
Does your CSM need to be revised?

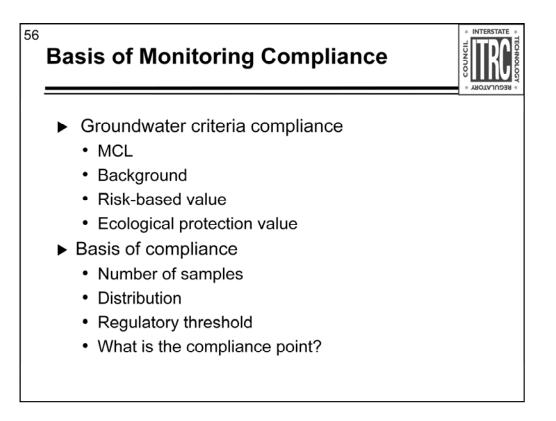


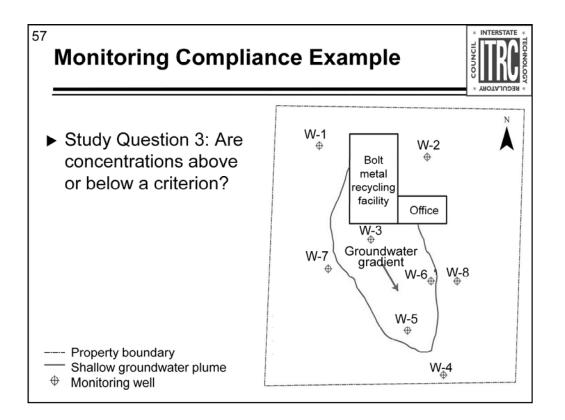


Two of the 10 questions are directly related to compliance.

Question 4 deals with when a particular criterion will be met, which is not considered in the IBT. Please refer to the web-based document for more information on Question 4.







Determined that wells W-1 and W-2 are background wells. Based on our CSM, wells W-7 and W-8 are side gradient.

For this question, focusing on wells W-3, W-4, W-5 and W-6, within the plume and/or downgradient of the suspected source.

# 58 Monitoring Compliance Dataset



► Is the site in compliance?
Are chemical concentrations
less than or greater than a
criterion?

- · Selection of data set
- Statistical analysis of data set
  - Methods and tools
- Interpretation of results
- Uncertainty

Arsenic (μg/l)								
Date	W-3	W-6	W-5	W-4				
1/2009	78	23	11.9	2.9				
4/2009	79	28	10.9	3.0				
7/2009	66	22	9.1	2.7				
10/2009	67	21	8.5	2.6				
1/2010	90	26	11.1	2.7				
4/2010	89	27	11.4	2.8				
7/2010	73	25	9.4	2.4				
10/2010	70	24	9.2	2.3				
1/2011	72	26	10.8	2.6				
EPA Drinking Water MCL = 10 μg/I								

Although W-3 and W-6 are several times the MCL and W-4 is below the MCL, you may think you might not need statistics. Misconception. There is always uncertainty in the results.

# <sup>59</sup> Monitoring Compliance Dataset



- Is the site in compliance? Are chemical concentrations less than or greater than a criterion?
  - · Selection of data set
  - Statistical analysis of data set
    - Methods and tools
  - Interpretation of results
  - Uncertainty

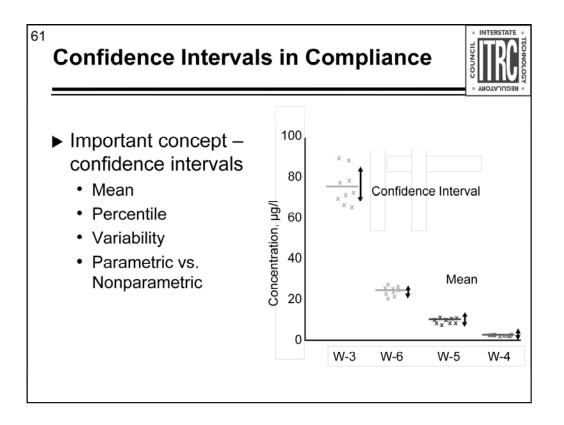
Arsenic (μg/l)								
Date	W-3	W-6	W-5	W-4				
1/2009	78	23	11.9	2.9				
4/2009	79	28	10.9	3.0				
7/2009	66	22	9.1	2.7				
10/2009	67	21	8.5	2.6				
1/2010	90	26	11.1	2.7				
4/2010	89	27	11.4	2.8				
7/2010	73	25	9.4	2.4				
10/2010	70	24	9.2	2.3				
1/2011	72	26	10.8	2.6				
EPA Drinking Water MCL = $10 \mu g/l$ Data are measured values								

# 60 Monitoring Compliance Dataset



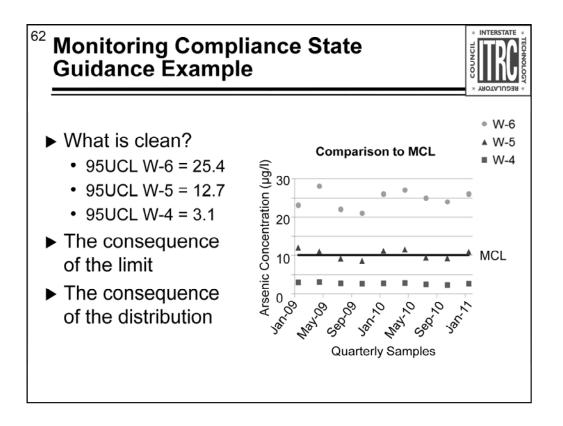
- Is the site in compliance? Are chemical concentrations less than or greater than a criterion?
  - · Selection of data set
  - Statistical analysis of data set
    - Methods and tools
  - Interpretation of results
  - Uncertainty

Arsenic (μg/l)								
Date	W-3	W-6	W-5	W-4				
1/2009	78	23	11.9	2.9				
4/2009	79	28	10.9	3.0				
7/2009	66	22	9.1	2.7				
10/2009	67	21	8.5	2.6				
1/2010	90	26	11.1	2.7				
4/2010	89	27	11.4	2.8				
7/2010	73	25	9.4	2.4				
10/2010	70	24	9.2	2.3				
1/2011	72	26	10.8	2.6				
EPA Drinking Water MCL = 10 µg/I								



Many kinds of confidence intervals

Mean (average) is a typical confidence measure.



This site is in the State of Washington, and we must use guidance that has been designed by the state. This is one method, and it doesn't match up with what is in the Unified Guidance. In the web-based document we identify the topic areas where programs are likely to have guidance (see Section 2.2.1)

What is considered clean is most-often determined by the regulatory program (under law, regulation or guidance). Those requirements may not have a statistical basis.

## <sup>63</sup> Monitoring Compliance State Guidance Example



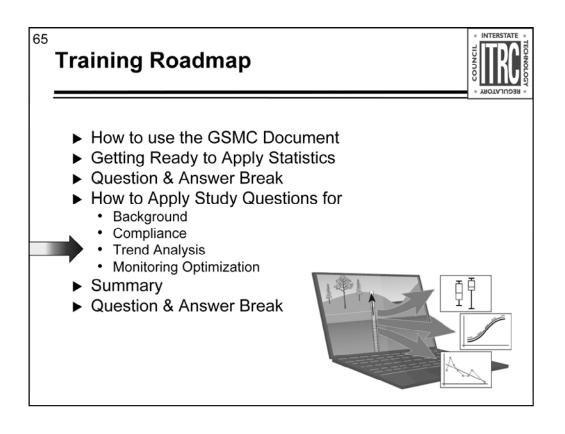
Arsenic (µg/l)	W-3	W-6	W-5	W-4		
SD	8.8	2.4	1.2	0.2		
Mean	76.0	24.7	10.3	2.7		
Confidence (0.05)	5.8	1.5	0.8	0.2		
UCL	83.8	25.4	12.7	3.1		
Max Data Value	90	28	11.9	3.0		
10% Data >MCL	Yes	Yes	Yes	No		
	OUT	OUT	OUT	IN		
Are 10% of the data > MCL?			MCL 10 μg/l			
			In = In Compliance			
		Out = Out of Compliance				
	SD Mean Confidence (0.05) UCL Max Data Value 10% Data >MCL	SD         8.8           Mean         76.0           Confidence (0.05)         5.8           UCL         83.8           Max Data Value         90           10% Data >MCL         Yes           OUT           ata	SD         8.8         2.4           Mean         76.0         24.7           Confidence (0.05)         5.8         1.5           UCL         83.8         25.4           Max Data Value         90         28           10% Data >MCL         Yes         Yes           OUT         OUT         OUT	SD         8.8         2.4         1.2           Mean         76.0         24.7         10.3           Confidence (0.05)         5.8         1.5         0.8           UCL         83.8         25.4         12.7           Max Data Value         90         28         11.9           10% Data >MCL         Yes         Yes         Yes           OUT         OUT         OUT         OUT           ata         MCL 1         In = In Compliance		

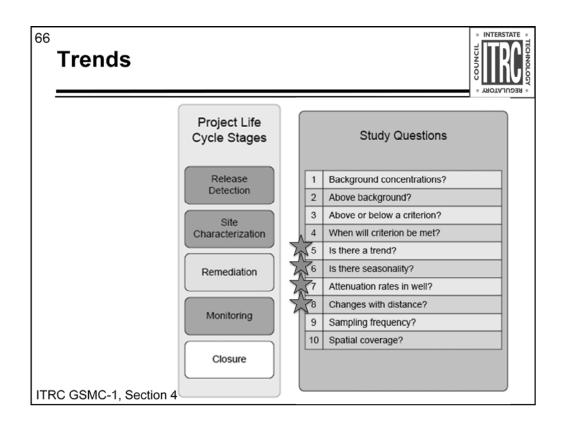
Data from wells W-3, W-6, W-5 and W-4 were evaluated using the State of Washington compliance test:

- Is the upper 95% CI on mean less than the cleanup standard?
- Is the maximum data point value less than 2x the cleanup standard?
- Are 10 Percent of the data above the cleanup standard?
  - 1. For W-3, that would be no, no and yes, so not clean
  - 2. For W-6, that would be no, no and yes, so not clean
  - 3. For W-5, that would be no, yes, and yes, so not clean
  - 4. For W-4, that would be yes, yes, and no, so clean

<sup>64</sup> Monitori Example										
Analysis	<ul> <li>40 CFR 264.99 evaluation using USEPA's Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities, Unified Guidance (2009)</li> </ul>									
	Arsenic (µg/l)	W-3	W-6	W-5	W-4					
	SD8.82.41.20.2Mean76.024.710.32.7									
	Confidence (0.05) 5.8 1.5 0.8 0.2									
	UCL	83.8	25.4	12.7	3.1					
	LCL	70.2	23.1	9.5	2.5					
		OUT	OUT	IN	IN					
	MCL 10 μg/l In = In Compliance Out = Out of Compliance									

As referenced in the web-based document, when looking at a RCRA situation (based on 40 CFR 264.99), the US EPA's Unified Guidance document is typically used.





Four of the 10 questions are directly related to trends. Trend analyses are indirectly related to all questions. For example, trend analyses are a component of exploratory data analysis. In addition, one also just consider when trends when establishing or comparing to background or how to optimize monitoring networks. Although trend analyses can be relevant to any of the project life cycle phases, they are most important to remediation, monitoring, and closure.

Question 8 deals with spatial trends, which are not considered in the IBT. Please refer to section 5.15 of the web-based guidance document for more information on spatial statistical methods.

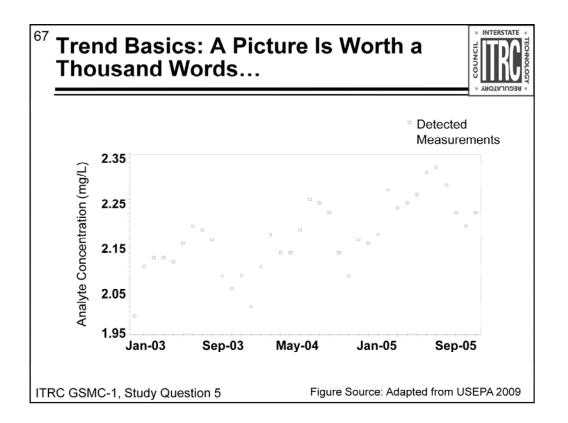


Figure Source: USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." In Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency. www.epa.gov

Start with a scatter plot of your data ... three years of data from a single well

### scatter plots

Graphical representation of multiple observations from a single point used to illustrate the relationship between two or more variables. An example would be concentrations of one chemical on the x-axis and a second chemical on the y-axis. They are a typical exploratory data analysis tool to identify linear versus nonlinear relationships between variables (Unified Guidance).

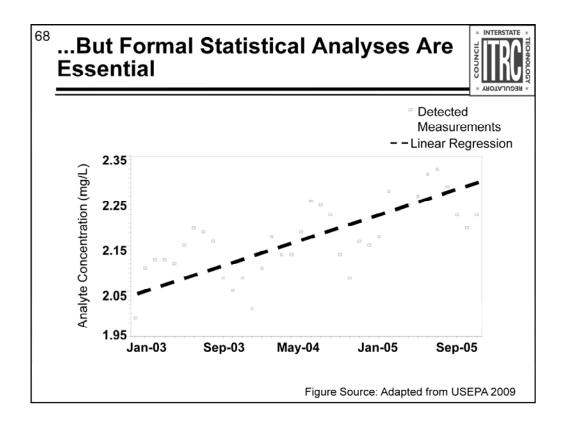


Figure Source: USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." In Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency. www.epa.gov

### Formal statistical tests answer:

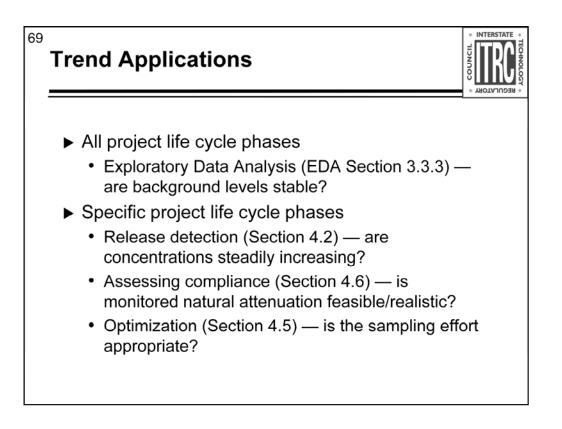
1) is there a trend (cyclical, increasing, decreasing)? and/or 2) what is the slope (or rate of change)?

### linear regression analysis

A parametric statistical method to measure the linear trend of a data set using data point regression residuals that are based on assumptions of normality, homoscedasticity, and independence (Unified Guidance).

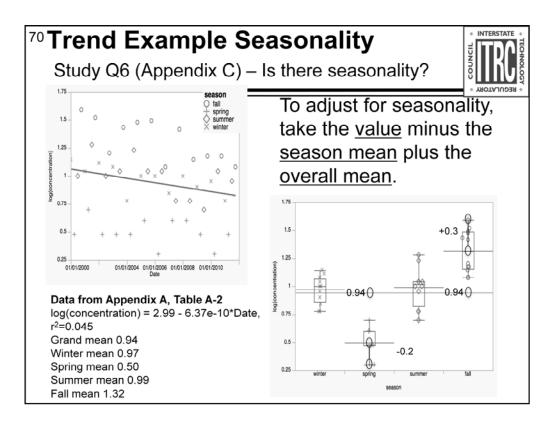
### parametric

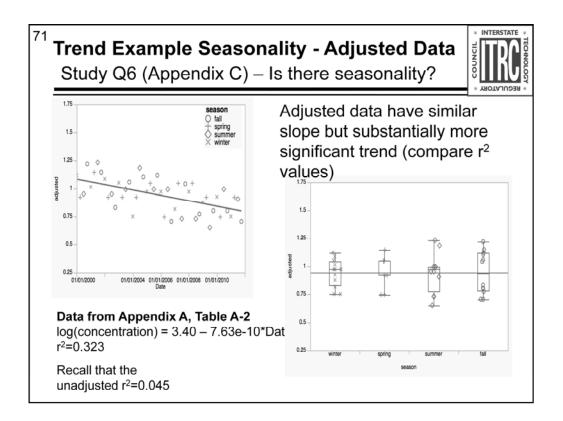
A statistical test that depends upon or assumes observations from a particular probability distribution or distributions (Unified Guidance).

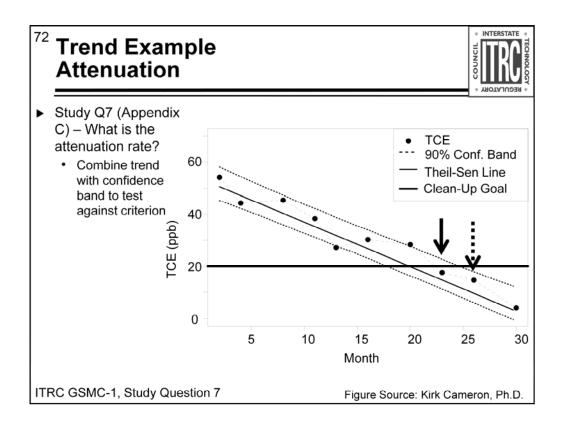


Exploratory data analysis applies to all project life cycle phases. Several statistical tests assume that concentrations are stable. For example, the background statistics assume there is no trend in the background well.

Seasonal or cyclical changes in concentrations can appear consistent with transient changes in concentrations. For documenting a release it is important to show changes are steadily increasing by using trend analyses and documenting statistical significance. Trend analyses can support assessment of monitored natural attenuation by estimating a rate of change or a date that a target concentration might be attained.







Data used to generate the figure are from the EPA's Unified Guidance, Example 21-7. USEPA. 2009. "Statistical Analysis of Groundwater Monitoring Data at RCRA Facilities." In Unified Guidance EPA 530/R-09-007. Washington DC: United States Environmental Protection Agency. www.epa.gov

Statistical confidence bands on observed trends can help to establish compliance. The scatterplot graph displays TCE concentrations over time at a monitoring well. Note that first sample result less than the criterion has an upper confidence band on the individual concentrations greater than the criterion. The upper confidence bound for the next result is less than the criterion. The confidence bands are calculated for the predicted values (y-axis, dependent variable, TCE concentration in the well).

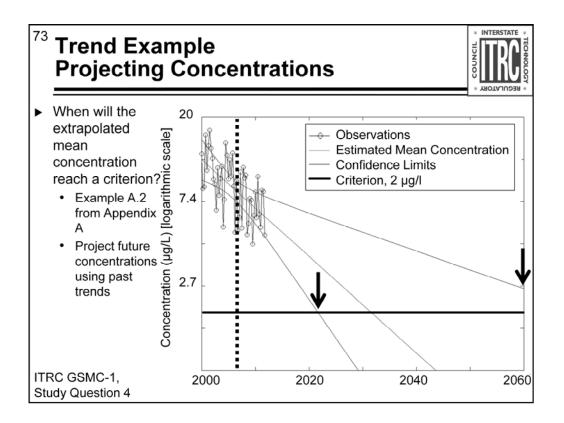
The Theil-Sen test is a nonparametric (no statistical distribution assumed) test that can be used with or without seasonality.

#### confidence interval

Statistical interval designed to bound the true value of a population parameter such as the mean or an upper percentile (Unified Guidance).

#### nonparametric

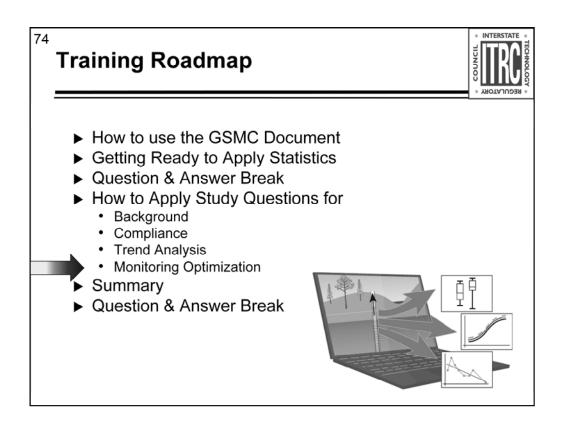
Statistical test that does not depend on knowledge of the distribution of the sampled population (Unified Guidance).

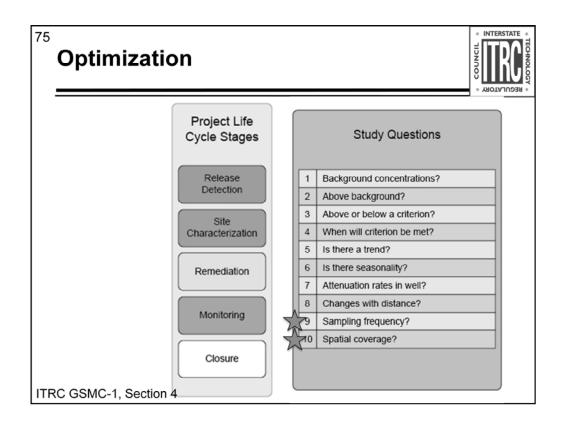


Statistical confidence bands on observed trend models can help to establish compliance. Note that the confidence bands on the model are quite different from those calculated for individual points shown on the previous slide. For this example the projected dates of compliance run from about 2020 to 2075 (off the x-axis scale).

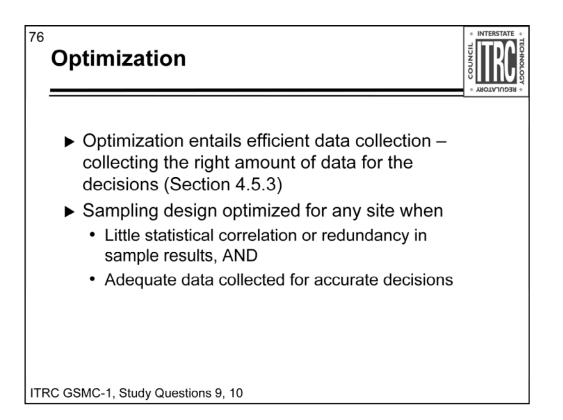
Extrapolating from current conditions to future mean concentrations leads to wider and wider confidence limits. This is due to extrapolating from the mean x or mean of dates sampled to future conditions. The mean data of the measured data (mean x) is marked by the dashed line. Confidence is greatest at the mean x-value (date) and expands with greater lags. Consider that when one extrapolates to possible future outcomes that there is a mean slope but also a range of possible slopes. We are also familiar with increasing bounds on model predictions based on predicted future storm tracks.

Figure A-11. How long until the compliance goal is met?





There are two study questions that deal specifically with optimization or determining the appropriate sampling frequency or well selection.



Similar results collected from nearby locations or closely-spaced times may indicate correlation or redundancy

Optimization can lead to either a larger or smaller number of wells, number of samples, or analytical suites

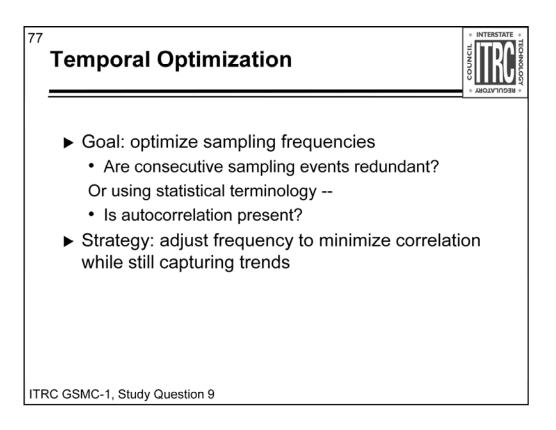
Sites at later stages in the project life cycle are more likely to have redundant data and the information necessary to conduct optimization

Optimization, like any statistical sampling procedure, should be completed with adequate site knowledge based on the conceptual site model.

Also, note there is a related ITRC team and guidance document http://gro-1.itrcweb.org

# **Geostatistics for Remediation Optimization**

Leads: Ning-Wu Chang (nchang@dtsc.ca.gov) and Harold Templin (htemplin@idem.in.gov) Project: Remediation optimization can improve performance, increase monitoring efficiency, and justify contaminated site decisions. For complex groundwater cleanup projects, however, optimizing the remediation performance monitoring system is challenging. A simple, deterministic decision flow chart may not adequately account for complex site conditions. Geostatistical approaches can be used as tools to evaluate optimization opportunities to improve groundwater remediation performance and monitoring. Geostatistical approaches for remediation optimization typically use spatial and temporal statistics to estimate correlations and redundancy between sampling locations and events. These approaches also identify areas and periods of high statistical uncertainty in a groundwater monitoring network over time. This project will develop a web-based guidance document and Internet-based training course to help state regulators and project managers understand how to use geostatistical approaches for remediation optimization optimization and make better decisions regarding site cleanup.



Phased sampling approach typical of many sites-

Generally sample more frequently in initial project phases

Reduce frequency once temporal relationships can be established

Common tests to identify temporal correlation (autocorrelation) include sample autocorrelation function (ACF) and rank von Neumann. The web-based document has information on methods to assess autocorrelation, including the example in Appendix A-2 that was discussed in the last section of this training.

autocorrelation

Correlation of values of a single variable data set over successive time intervals (Unified Guidance).

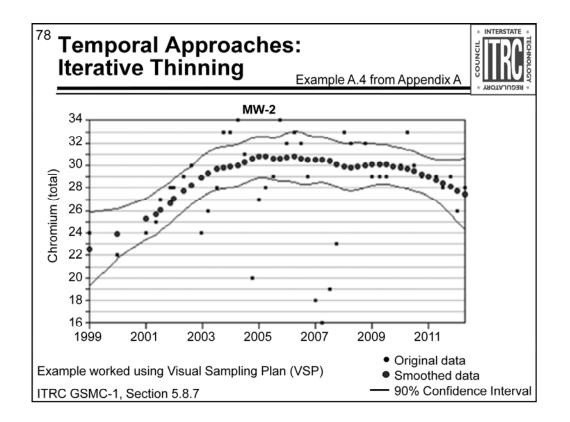


Figure A-15. Visual Sampling Plan (VSP) results for concentration (mg/l) from MW-2.

**Iterative thinning method**. Iterative thinning examines whether sampling frequencies can be reduced due to temporal redundancy in the sampling events. This approach identifies redundancy by first estimating a baseline trend using the full data set, after which the trend is repeatedly re-estimated using subsets of the full data to identify the average number of data points needed to accurately reconstruct the baseline. The computations in iterative thinning create a series of 'what if' scenarios estimating the nature of the trend that would have been identified if only some of the existing data had been sampled. The overriding principle in iterative thinning is that if a trend can be accurately reconstructed using fewer sampling events, the optimal sampling frequency should be based on this smaller number.

This example is from Appendix A, A-4 and used the VSP software (see Appendix D.23). The black symbols represent the original quarterly well sample data and the red symbols are the smoothed trend using all data. The blue band is the 90% confidence interval on the trend. According to the VSP output, the optimal sampling frequency for MW-2 would be 227 days. If a semi-annual frequency (180 days) were proposed for future sampling in this well, there would be a 50% reduction in sampling costs with no significant difference in the ability to monitor trends in this well.

The user selects the following **Simulation Parameters** in VSP: Smoothing bandwidth: for example 0.3 Number of simulations: for example 500 Confidence interval (CI) confidence level: 90% [for the smoothed line] Percent of simulated trend required to be within the CI: for example 75%

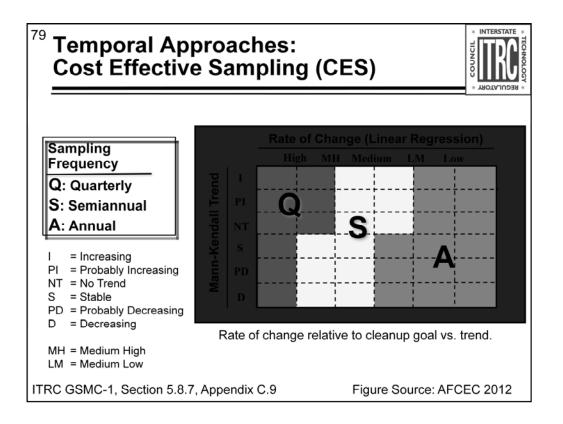
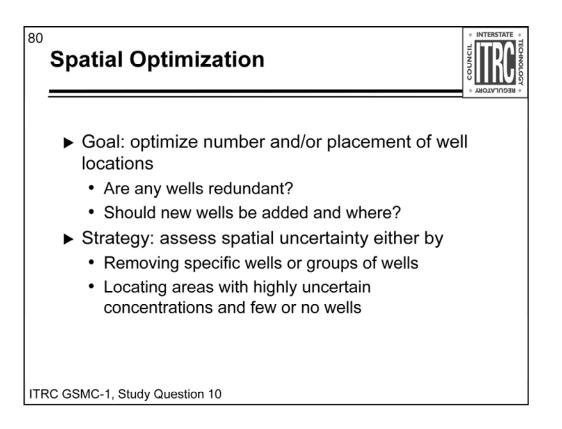


Figure Source: Air Force Civil Engineer Center (AFCEC). 2012. "Monitoring and Remediation Optimization System (MAROS) Software, User's Guide and Technical Manual." In: Air Force Center for Environmental Excellence. http://www.gsi-net.com/en/software/free-software/maros-30.html.

See Also Ridley, M.N., V.M. Johnson, and R.C. Tuckfield. 1995. Cost-Effective Sampling of Groundwater Monitoring Wells. Vol. UCRL-JC-118909. Livermore, CA: Lawrence Livermore National Laboratory

**Cost-effective sampling method (CES)**. In CES, a linear trend is estimated for each chemical-well pair and then classified according to the slope of the apparent trend as well as how much variation exists around the trend. Trends with relatively 'flat' slopes (small rates of change) and low variation are recommended for less frequent sampling, while trends with higher slopes or higher degrees of variation are targeted for more frequent sampling. The overriding principle is to (1) sample more frequently at locations where the apparent changes are more dynamic and associated with the greatest statistical uncertainty, and (2) sample less frequently when the trend is changing little and is statistically more certain (that is, less variable). A high rate is change is 2 times the cleanup goal, medium is equal to the cleanup goal, and low is 50% of the cleanup goal.



Spatial optimization is a challenging objective and an active area for research

Requires lots of data; broad spatial coverage

Optimization software results should be checked against what is known or hypothesized about contamination via the CSM (see section 3.2 of the web-based document), and other lines of evidence.

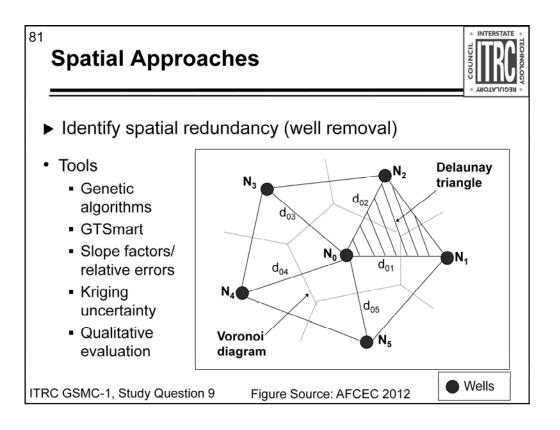


Figure Source: Air Force Civil Engineer Center (AFCEC). 2012. "Monitoring and Remediation Optimization System (MAROS) Software, User's Guide and Technical Manual." In: Air Force Center for Environmental Excellence. http://www.gsi-net.com/en/software/free-software/maros-30.html.

**Spatial redundancies** in a monitoring network can be identified, leading to fewer sampling points and more cost-effective monitoring. In this case, multiple tools and approaches exist to answer the question. For instance, nearest neighbor estimation can be combined with leave-one-out cross-validation to generate the slope factors in MAROS. Some approaches remove one well at a time or groups of wells, then try to either reproduce plume maps using the reduced network or minimize the statistical uncertainty of the resulting network

3TMO uses qualitative evaluation (see Appendix D.1); GTS uses quasi-genetic algorithm called GTSmart (see Appendix D.6); Summit Tools uses genetic algorithm (see Appendix D.21); MAROS uses slope factors (see Appendix D.11); VSP uses kriging uncertainty (see Appendix D.23)

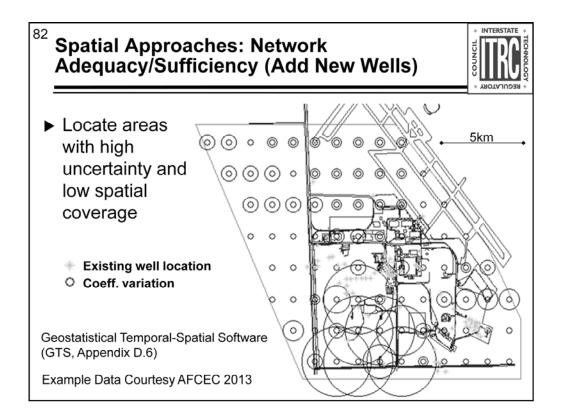
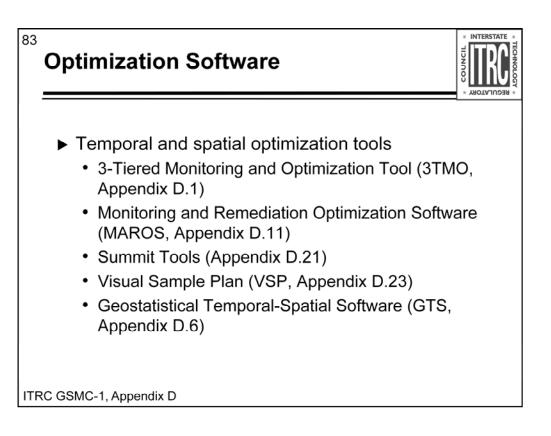


Figure Source: Map of network adequacy evaluation results developed in GTS using example data from the software. Example data courtesy AFCEC 2013.

The plus symbols locate existing wells. The circles represent current estimates of uncertainty in concentrations. Add wells where uncertainty is high and coverage of existing wells is low. Remove wells is coverage is high and uncertainty is low. The CSM should be considered when making decisions to either add or remove wells from the monitoring network.



Spatial optimization generally relies on geostatistical methods to understand the distance over which sample results are correlated and the strength of such relationships. Temporal optimization relies on estimating trends and the uncertainty associated with those trends.

#### 3TMO

The 3-Tiered approach to LTMO is unique when compared to existing LTMO statistical applications because it focuses on qualitative factors that are supported by quantitative statistical analysis. The spatial analysis included in 3TMO is a qualitative evaluation facilitated by the Map Tool; the spatial importance of each well is not quantitatively determined using geostatistics.

#### MAROS

Uses simple statistics and decision frameworks to prioritize data collection efforts and link data to defensible site management decisions. Can use results from this software to develop lines of evidence, which combined with professional judgment, can be used to inform site management decisions for safe and economical long-term monitoring of groundwater plumes. Can use MAROS to help design and calculate remediation performance metrics and as a tool to evaluate progress toward site remedial goals.

### GTS

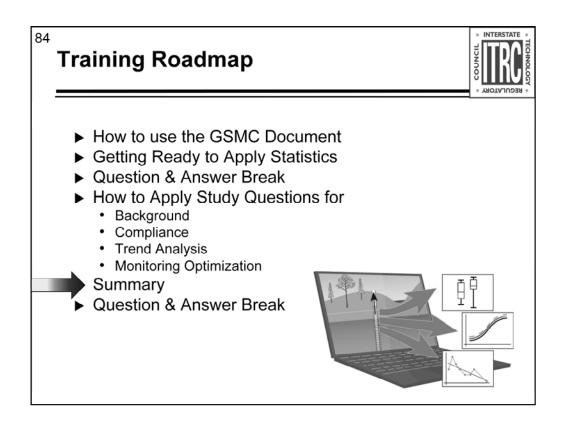
GTS has five modular components linked together in a user-friendly interface: Prepare, Explore, Baseline, Optimize, and Predict. The Optimize component runs two distinct types of temporal optimization—iterative thinning and temporal variograms—as well as spatial optimization involving both a search for statistical redundancy and an assessment as to whether and where new wells should be added. Finally, the Predict module focuses on flagging newly imported data that are inconsistent with projected trends and maps.

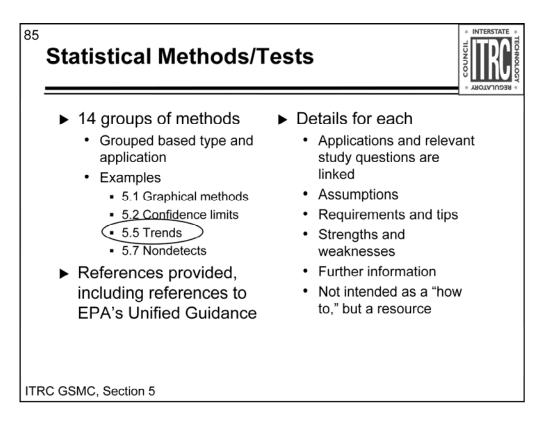
#### Summit Tools

SampleOptimizer and SampleTracker are tools for both spatial and spatio-temporal analysis for monitoring network optimization. SampleOptimizer applies mathematical optimization to monitoring networks in an easy-to-use desktop software tool. SampleTracker reviews new monitoring data against historical data.

# VSP

VSP has a module to evaluate redundancy of wells. The well redundancy modules in VSP can identify redundant wells and identify a technically defensible temporal spacing of observations for wells. The redundant well module uses a geospatial analysis based on kriging. VSP also has a module to help with new well placement to reduce estimation uncertainty. The sampling frequency well evaluation is applied on a well-by-well basis and is based on iterative thinning methods.

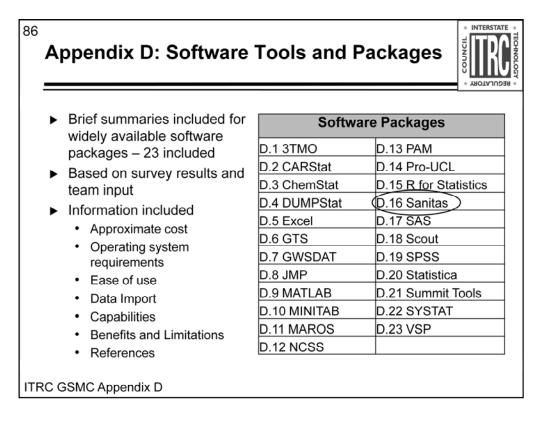




In Section 5 of the web-based document the statistical methods are grouped by their application.

There are references provided for the methods.

A lot of practical information about the methods is provided in Section 5.



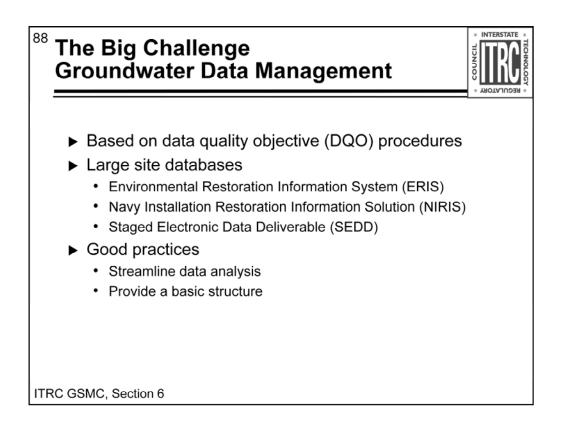
Appendix D includes summaries of 23 statistical software packages. Table D-1.

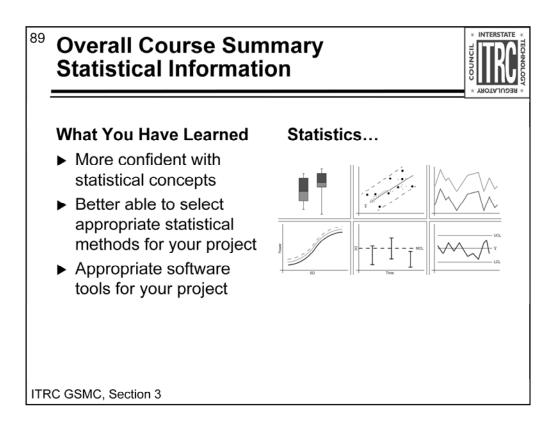
Each summary includes some practical information such as the approximate cost and capabilities.

You should be sure to carefully review any software to be sure that it is applicable to your needs.

Appendix D: Software Packages		
Statistical Method	Capability As Is	Capability with Scripts/Add-Ins
Handling of NDs		
Simple Substitution	•	N/A
Kaplan-Meier ROS		N/A N/A
Cohen/MLE	÷	N/A
Exploratory/Diagnostic Tools		
Summary Statistics	•	N/A
Distributional tests	÷	N/A
Outlier tests	•	N/A
Data Transformations		N/A
C GSMC Appendix D		

Ratings are none, some capability and full capability. These can be used to determine which software can implement the functions that you are looking for.

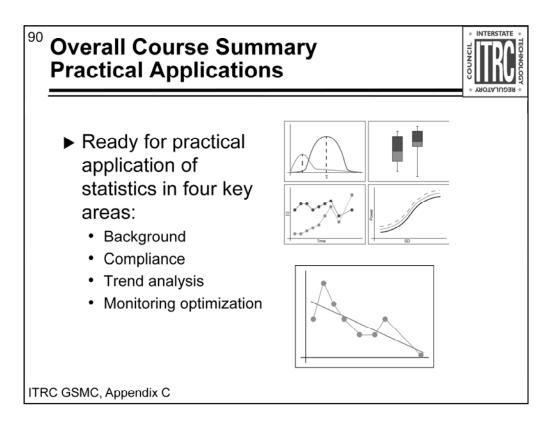


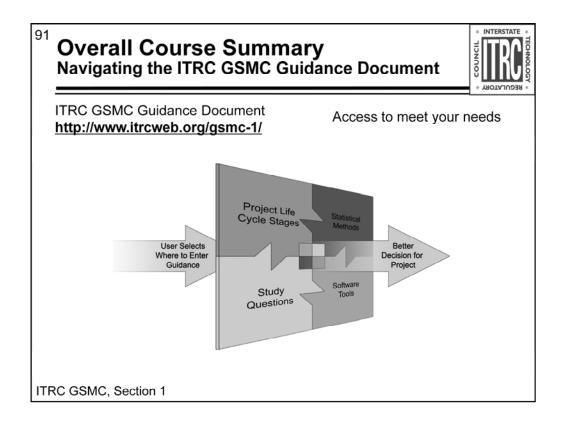


Statistical concepts

Select appropriate methods and software tools

Avoid misapplications of statistics at your groundwater sites.

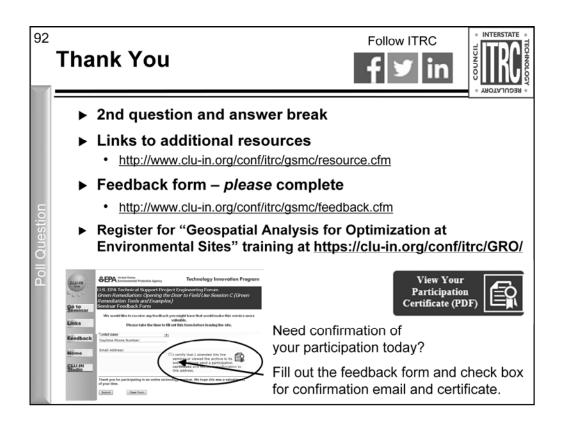




You learned how the Web-based guidance document fits together to give you a resource in evaluating or designing a statistical approach.

This document guides the project manager in using appropriate statistical methods to address common project tasks (such as evaluating whether a groundwater remedy is functioning effectively or whether there is a downward trend which supports a natural attenuation remedy selection) and make better decisions for their projects.

You may also find the GSMC Team as a resource for more information (Appendix G).



Links to additional resources: http://www.clu-in.org/conf/itrc/gsmc/resource.cfm

Your feedback is important – please fill out the form at: http://www.clu-in.org/conf/itrc/gsmc/feedback.cfm

# The benefits that ITRC offers to state regulators and technology developers, vendors, and consultants include:

✓ Helping regulators build their knowledge base and raise their confidence about new environmental technologies

- ✓ Helping regulators save time and money when evaluating environmental technologies
- $\checkmark$  Guiding technology developers in the collection of performance data to satisfy the requirements of multiple states

 $\checkmark$  Helping technology vendors avoid the time and expense of conducting duplicative and costly demonstrations

✓ Providing a reliable network among members of the environmental community to focus on innovative environmental technologies

# How you can get involved with ITRC:

 $\checkmark$  Join an ITRC Team – with just 10% of your time you can have a positive impact on the regulatory process and acceptance of innovative technologies and approaches

- ✓ Sponsor ITRC's technical team and other activities
- ✓Use ITRC products and attend training courses
- ✓ Submit proposals for new technical teams and projects