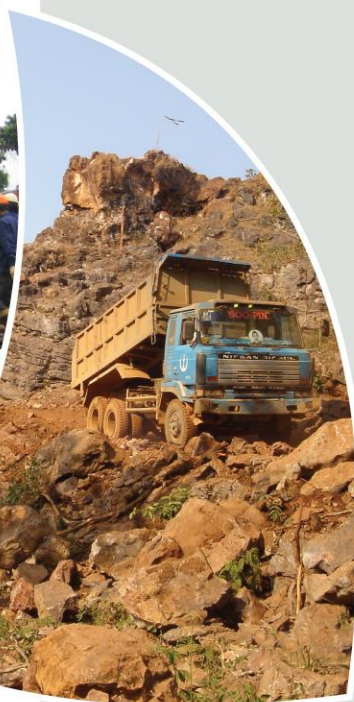




CSA Global
Resource Industry Consultants

Data Quality for Reporting – How Good is Good Enough?

Dennis Arne, PGeo (BC), RPGeo (AIG)
Principal Consultant - Geochemistry

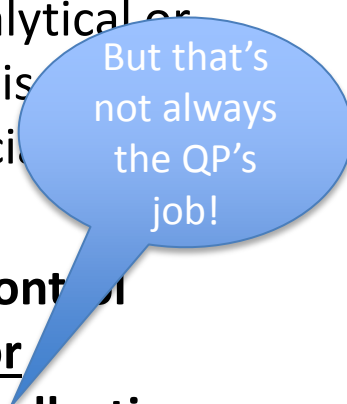


Outline

- What do the codes say?
 - NI43-101
 - JORC
- Sampling
- QAQC
- What should our quality expectations be?

NI43-101

- **Item 11: Sample Preparation, Analyses, and Security – Describe**
 - (a) **sample preparation methods and quality control measures employed before dispatch of samples** to an analytical or testing laboratory, the method or process of sample splitting and reduction, and the security measures taken to ensure the validity and integrity of samples taken;
 - (b) relevant information regarding sample preparation, assaying and analytical procedures used, the name and location of the analytical or testing laboratories, the relationship of the laboratory to the issuer, whether the laboratories are certified by any standards association, and the particulars of any certification;
 - (c) **a summary of the nature, extent, and results of quality control procedures employed and quality assurance actions taken or recommended to provide adequate confidence in the data collection and processing**; and
 - (d) the author's opinion on the adequacy of sample preparation, security, and analytical procedures.



But that's not always the QP's job!

JORC 2012 Table 1

- “The nature, quality and appropriateness of the assaying and laboratory procedures used and whether the technique is considered partial or total.”
- “For geophysical tools, spectrometers, handheld XRF instruments, etc, the parameters used in determining the analysis including instrument make and model, reading times, calibrations factors applied and their derivation, etc.”
- “Nature of quality control procedures adopted (standards, blanks, duplicates, external laboratory checks) and whether acceptable levels of accuracy (ie lack of bias) and precision have been established.”

Sampling & Analysis

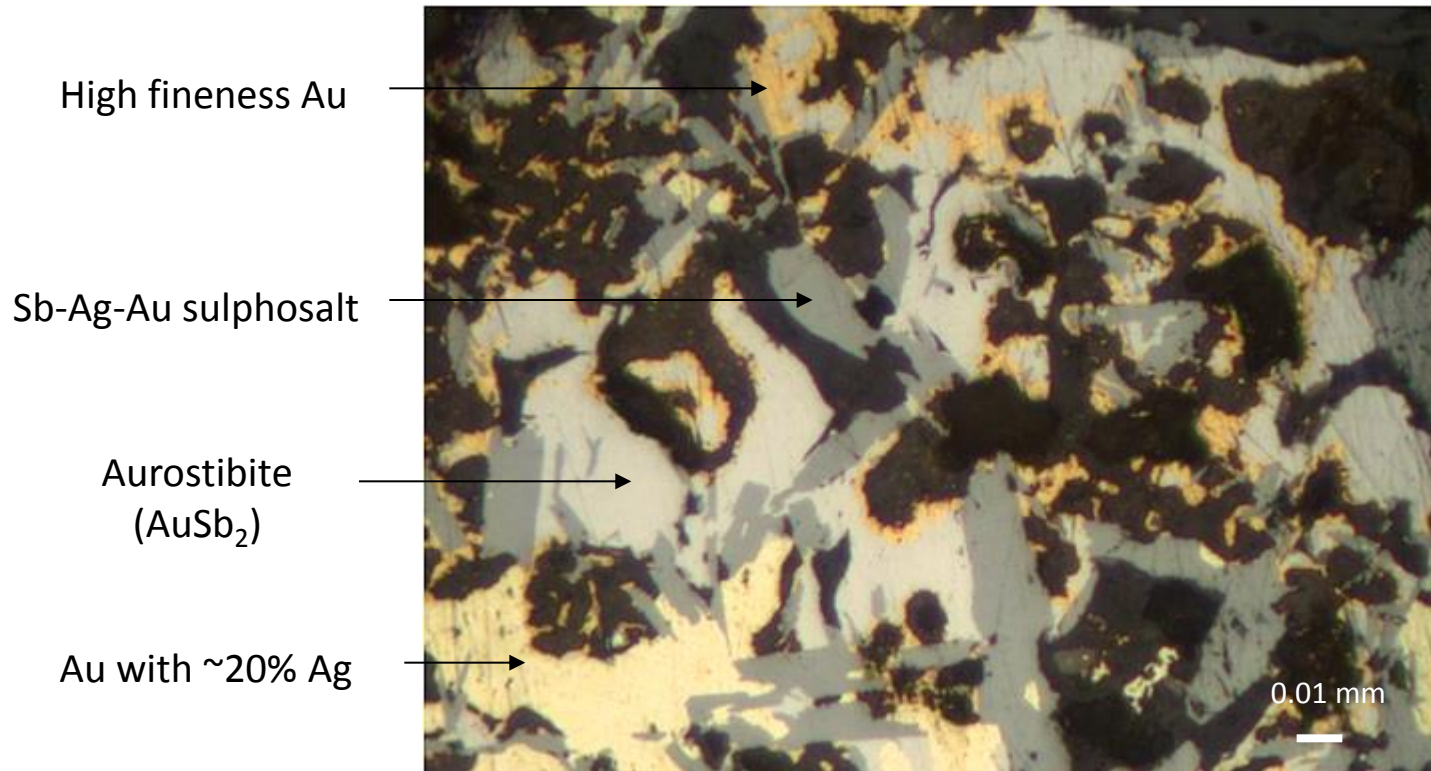
- Sample size, grain size and analytical methods generally default to convention.
- Should be optimised for individual projects.
- Objective is to obtain representative samples.
- The data should be relevant for proposed processing.
- Have we maximized precision (i.e. minimized uncertainty)?

Is Fire Assay Alone Appropriate?

CL Test	Oxidation	Grain Size (μm)	Grade	Au Recovery (%)
Oxide	100 %	96	7.0	96.0
Oxide	100 %	99	7.0	96.5
Transitional	95 %	108	3.4	90.7
Transitional	95 %	114	3.4	91.5
Fresh	0 %	90	4.4	2.0
Fresh	0 %	35	4.4	2.8
Fresh	0 %	20	4.4	5.3

- ~50 % of resource is from transitional material.
- Primary sulphide mineralization strongly refractory.
- Recoveries in the transition zone proportional to oxidation.
- Cyanide Au recoveries important for heap leach operation.

Effects of Complex Mineralogy



A fire assay gives a total Au value for this sample.

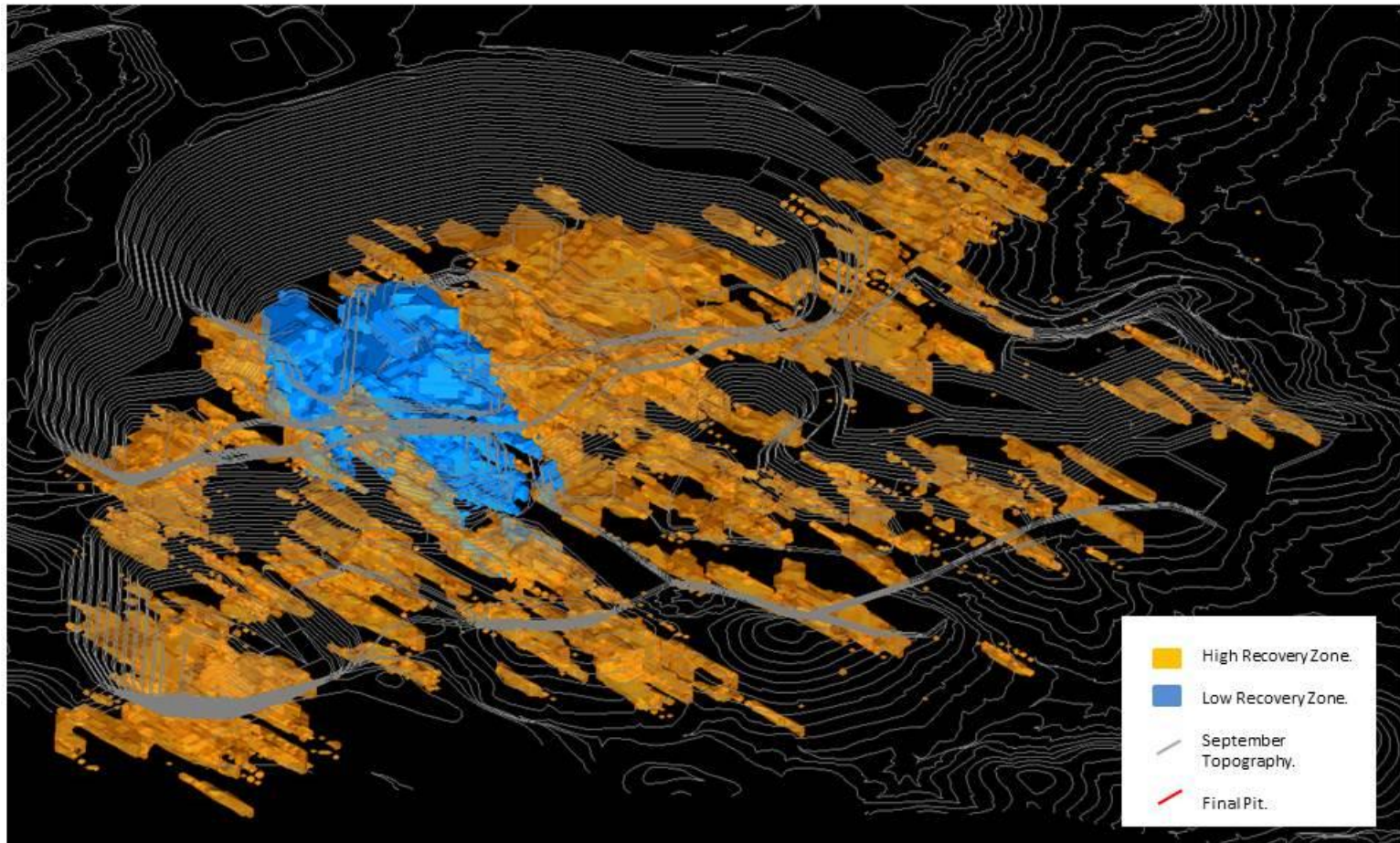
Appropriate Assay Methods

Sample	25 g Fire Assay (ppm)	25 g Fire Assay Repeat (ppm)	200 g Leachwell (ppm)	25 g Leachwell Tails Fire Assay (ppm)	Total Au from BLEG + Fire Assay (ppm)	% Cyanide soluble to total Au
Sample008	13.16	n/a	1.16	12.73	13.89	8.8
Sample009	17.23	13.03	5.00	12.32	17.32	33.0
Sample010	5.23	n/a	0.43	5.35	5.78	8.2
Sample011	13.12	12.33	4.72	5.50	10.22	37.0
Sample012	30.24	30.60	19.95	12.38	32.33	65.6
Sample013	38.70	27.98	47.85	0.75	48.6	98

Leachwell analysis with a tails fire assay of provides:

- an estimate of the CN-extractable Au in the sample
 - a bulk analysis – more precise

El Castillo Gold Recovery Block Model



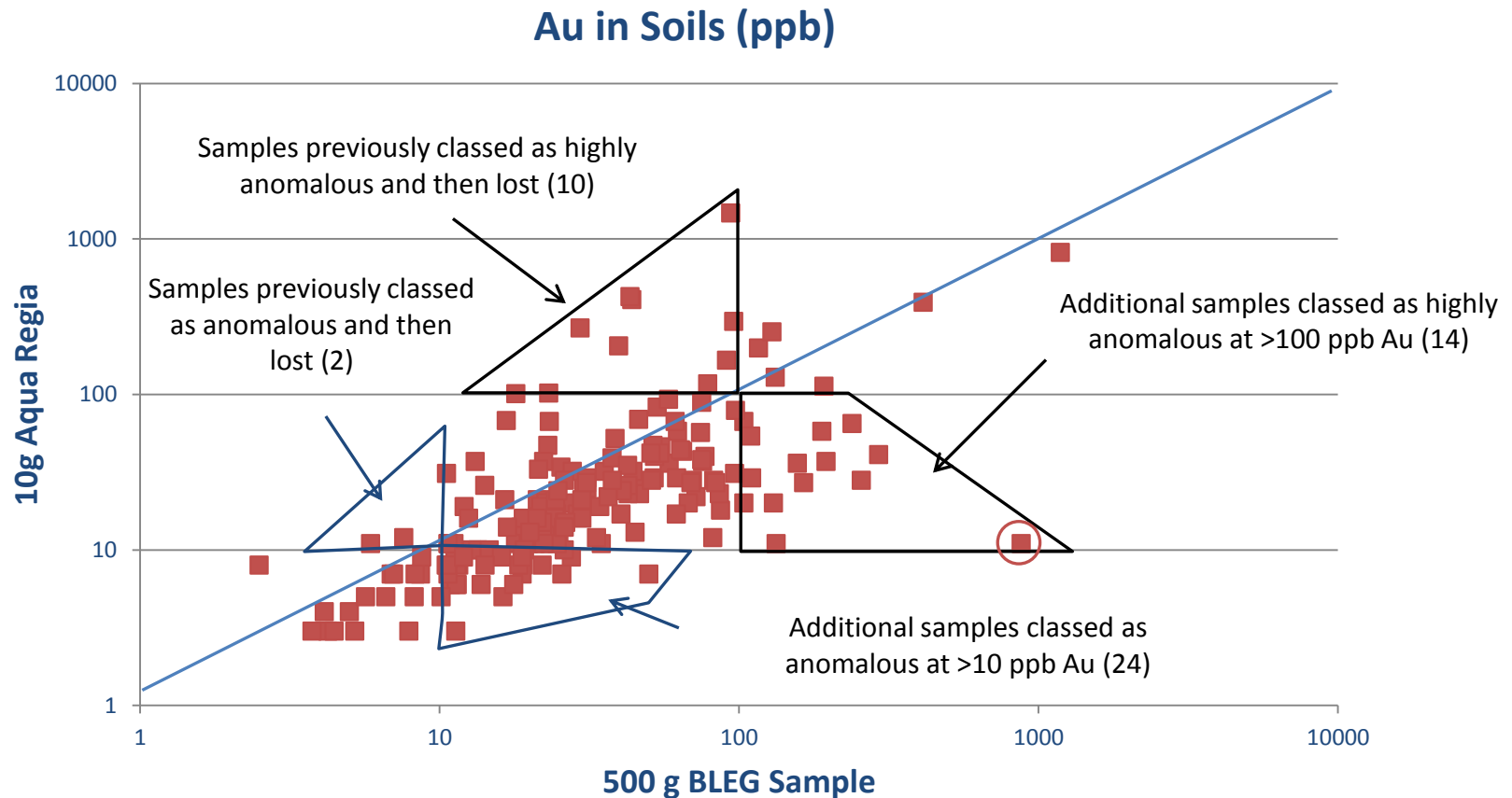
Either quantitative or semi-quantitative information on metal recoveries can be built into block models

Appropriate Sample Size?

	25 g FA	CV	LW + tails	SFA	Bulk CV (SFA/LW)
Average grade	13.23		13.73	13.11	
Average (all)		28			9
Average (area 1)		25			11
Average (area 2)		33			8
Average (> 10 ppm)		30			10
Average (> 5 ppm)		30			10

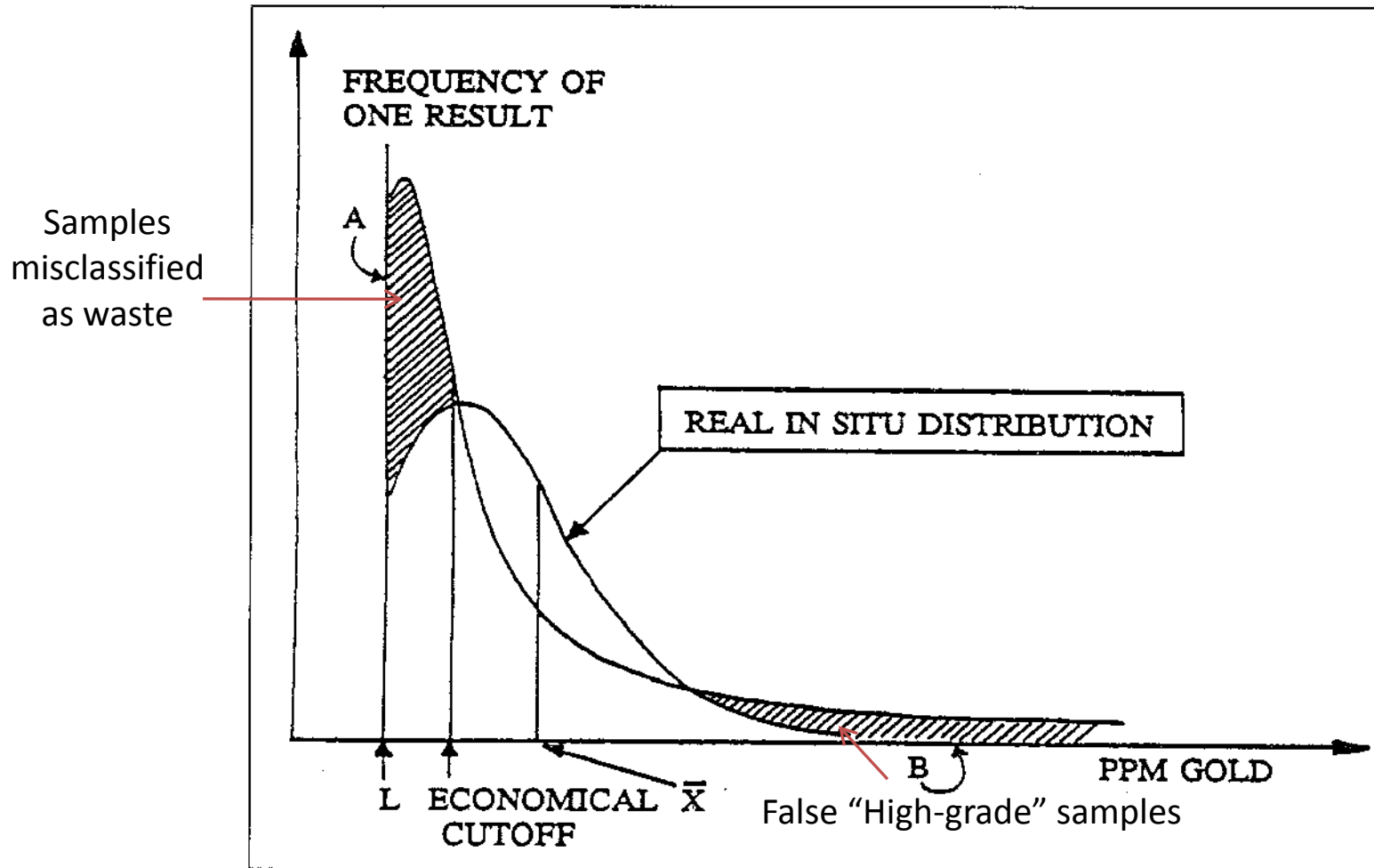
- Average grades and standard deviations calculated for 25 samples
- The bulk coefficient of variation is $\sim 1/3$ of that obtained for the 25 g fire assays

Particulate Au in Soil – Lost Opportunities?



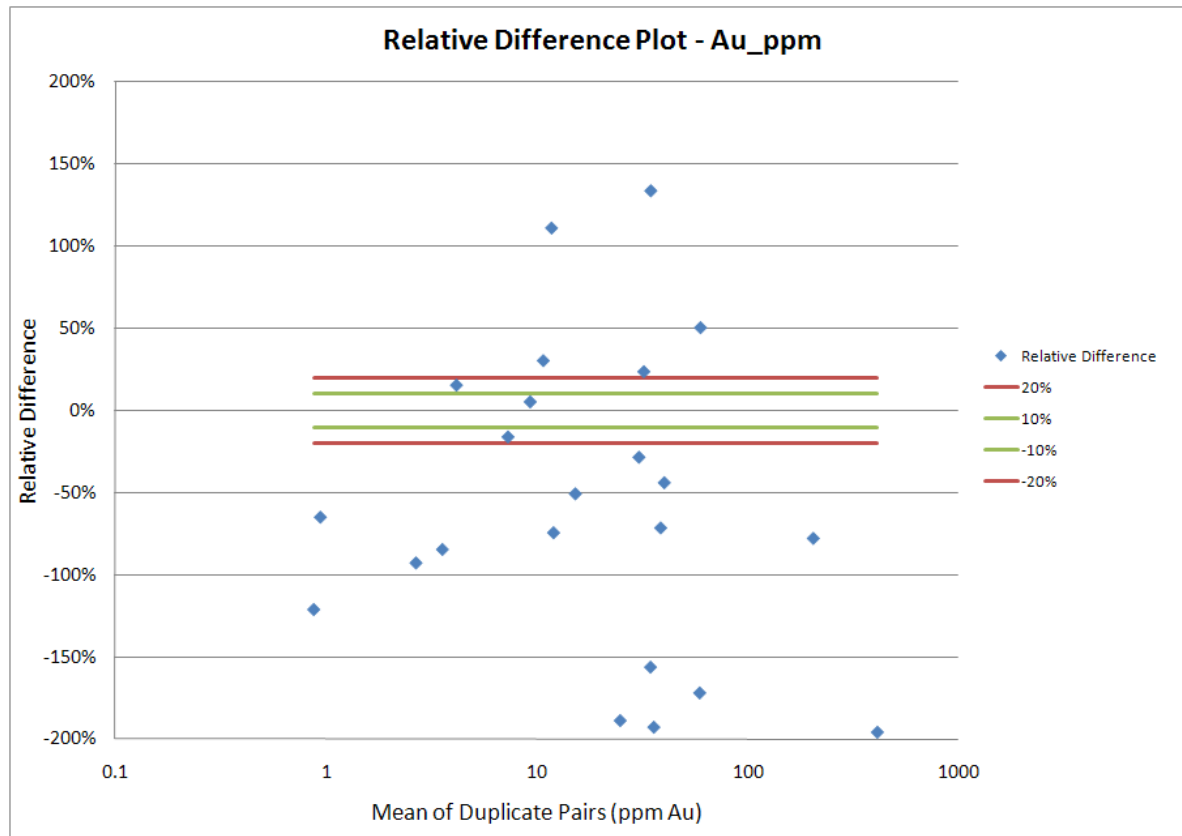
- Significantly more anomalous samples @ 10 ppb Au
- 20% more highly anomalous samples @ 100 ppb Au

Economic Consequences of Poor Sampling



Non-representative sampling does not reflect the underlying distribution.

Gold Assay Reproducibility Problems



- Relative differences between $\frac{1}{2}$ core and $\frac{1}{4}$ core check samples.
 - Bias toward higher values in the original half-core sample.

Upgrade in Gold Assay Results

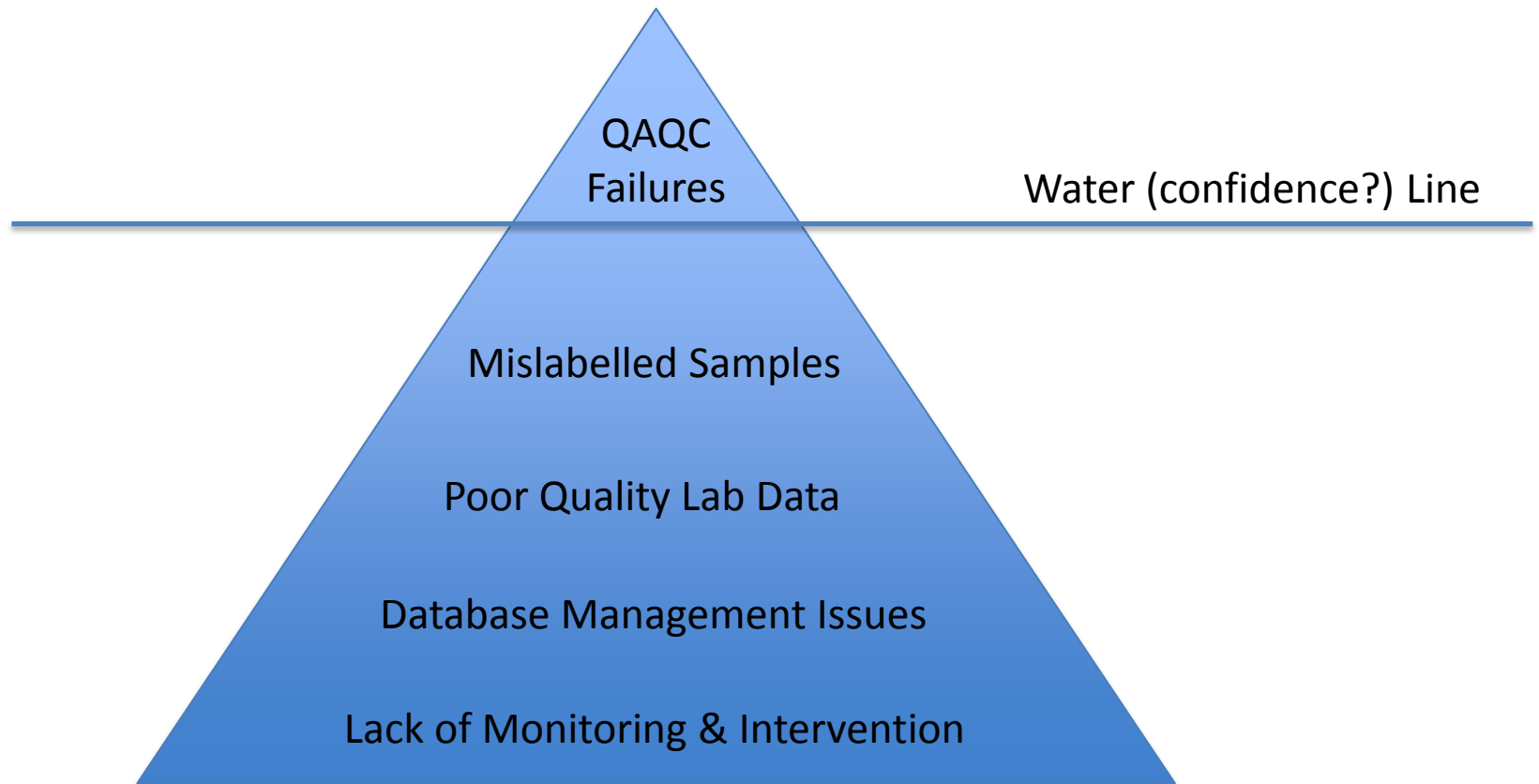
From (m)	To (m)	Interval (m)	Au (g/t) original fire assay	Au (g/t) metallic screen	Change %
116.0	117.5	1.5	2.26	2.35	4.1
117.5	119.0	1.5	47.9	44.7	-6.6
119.0	120.5	1.5	12.8	13.5	5.4
120.5	122.0	1.5	338	382	13.1
122.0	123.5	1.5	236	231	-2.3
123.5	125.0	1.5	73.0	92.8	27.1
125.0	126.5	1.5	681	880	29.3
126.5	128.0	1.5	17.2	16.2	-5.6
128.0	129.5	1.5	140	157	11.8
129.5	131.0	1.5	3.89	3.63	-6.6
Total: 12.0 metres (117.5m-129.5m) averaging:			193 g/t	227 g/t	17.50%

- Example is from a Au-rich epithermal system in Turkey
- Samples were re-assayed using 500 g screened metallics

Aspects of Setting up a QAQC Program

- What level of control is required?
- Where is the major source of variation (sampling?)
- Are the certified reference materials (CRMs) appropriate?
- Do the CRMs cover the range of expected grades & anticipated cut-off?
- Are suitable procedures documented?
- Can you access internal laboratory QAQC data?
- What constitutes a QAQC failure?

Tip of the Iceberg



Too many QAQC failures reduce confidence in the data

What CRM Value do I Use?

Table 14. Performance gates for precious and base metals by aqua regia digest ICPOES/MS.

Constituent	1 σ		2 σ		3 σ		5%	
	Low	High	Low	High	Low	High	Low	High
Antimony, Sb (ppm)	0.27	0.36	0.22	0.41	0.17	0.45	0.30	0.33
Arsenic, As (ppm)	2.1	3.8	1.3	4.7	0.4	5.6	2.8	3.1
Bismuth, Bi (ppm)	0.16	0.19	0.15	0.20	0.14	0.22	0.17	0.19
Cadmium, Cd (ppm)	0.10	0.11	0.09	0.12	0.08	0.13	0.10	0.11
Chromium, Cr (ppm)	614	719	561	772	509	825	633	700
Cobalt, Co (ppm)	71.5	76.1	69.3	78.3	67.0	80.6	70.1	77.5
Copper, Cu (ppm)	414	484	378	519	343	554	426	471
Gold, Au (ppb)	28	35	24	38	21	41	30	33
Lead, Pb (ppm)	20	23	18	24	17	25	20	22
Nickel, Ni (ppm)	187	208	176	218	166	229	187	207
Palladium, Pd (ppb)	28	42	22	49	15	55	33	37
Platinum, Pt (ppb)	40	55	32	63	24	71	45	50
Silver, Ag (ppm)	0.19	0.21	0.17	0.23	0.16	0.24	0.19	0.21
Zinc, Zn (ppm)	165	181	157	190	149	198	165	182

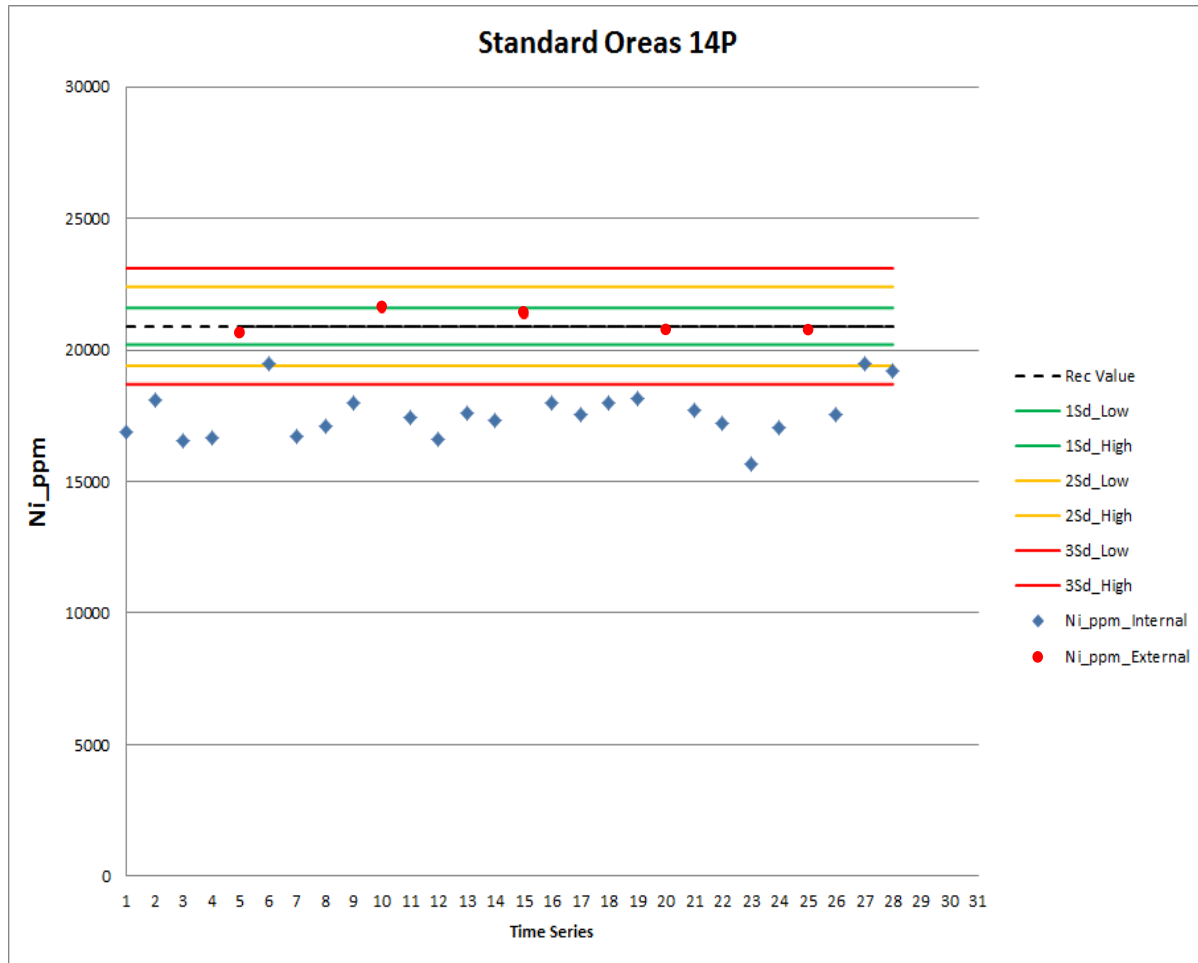
- Use the standard deviation from the clean round robin data
- Not the confidence interval on the mean determination
 - Make sure digestion method is appropriate

The CRM Should be Matrix Appropriate

- Inappropriate matrix
- Inappropriate grade
- Target anomaly >7 ppb Au
- Data needed to be cleaned

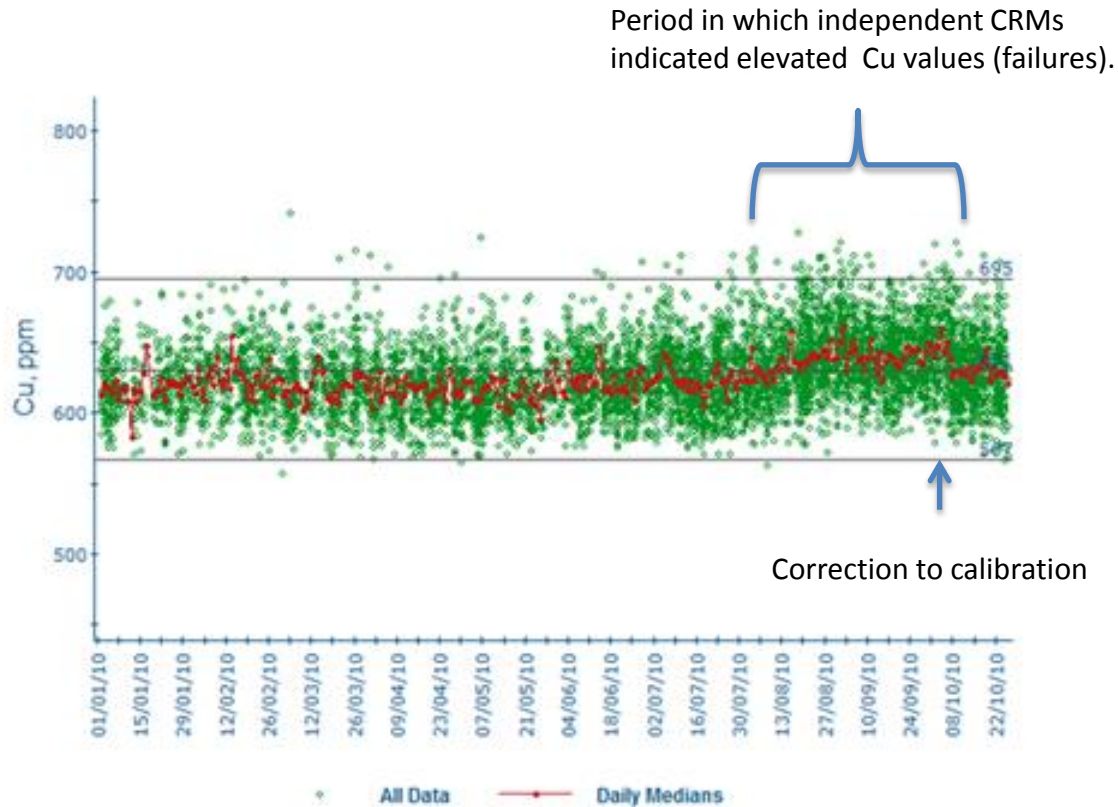
Sample Type	Au (ppb)
Soil	1.5
Soil	2
Fresh sulphide CRM	2720
Soil (memory effect)	32.1
Soil (memory effect)	7.5
Soil	3.4
Soil	2.5
Soil	2.3
Soil	2.2
Soil	1.5
Soil	1.3

Are the Sulphide CRMs Fresh?



Old bulk CRM (blue) compared to fresh CRM (red).

All Labs Have a Bias



- Internal laboratory QAQC data
- Re-assays performed after correction
- Intervention was timely – overall data quality >98 % pass rate

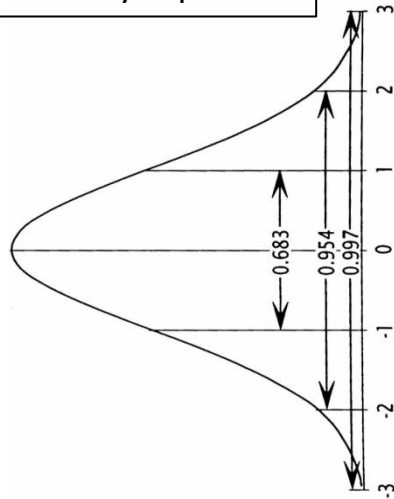
What Constitutes a CRM Failure?

Rule	Description	Chart Example
Rule 1	Any single data point falls outside the 3σ limit from the centerline (i.e., any point that falls outside Zone A, beyond either the upper or lower control limit)	<p>Rule 1: Any point beyond Zone A</p>
Rule 2	Two out of three consecutive points fall beyond the 2σ limit (in zone A or beyond), on the same side of the centerline	<p>Rule 2: two out of three consecutive points fall Zone A or beyond</p>

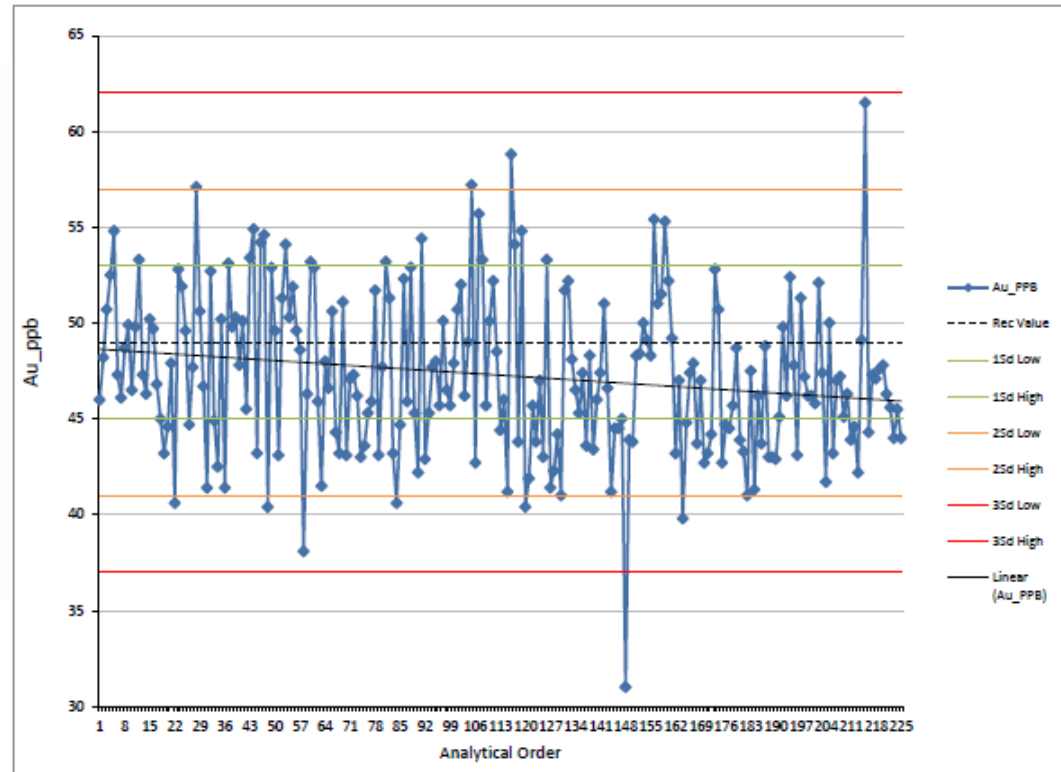
Codified in 1956 by the Western Electric Company

There Are Statistical “Failures”

11 of 225 > 2 SD
= ~5%, which is
statistically expected

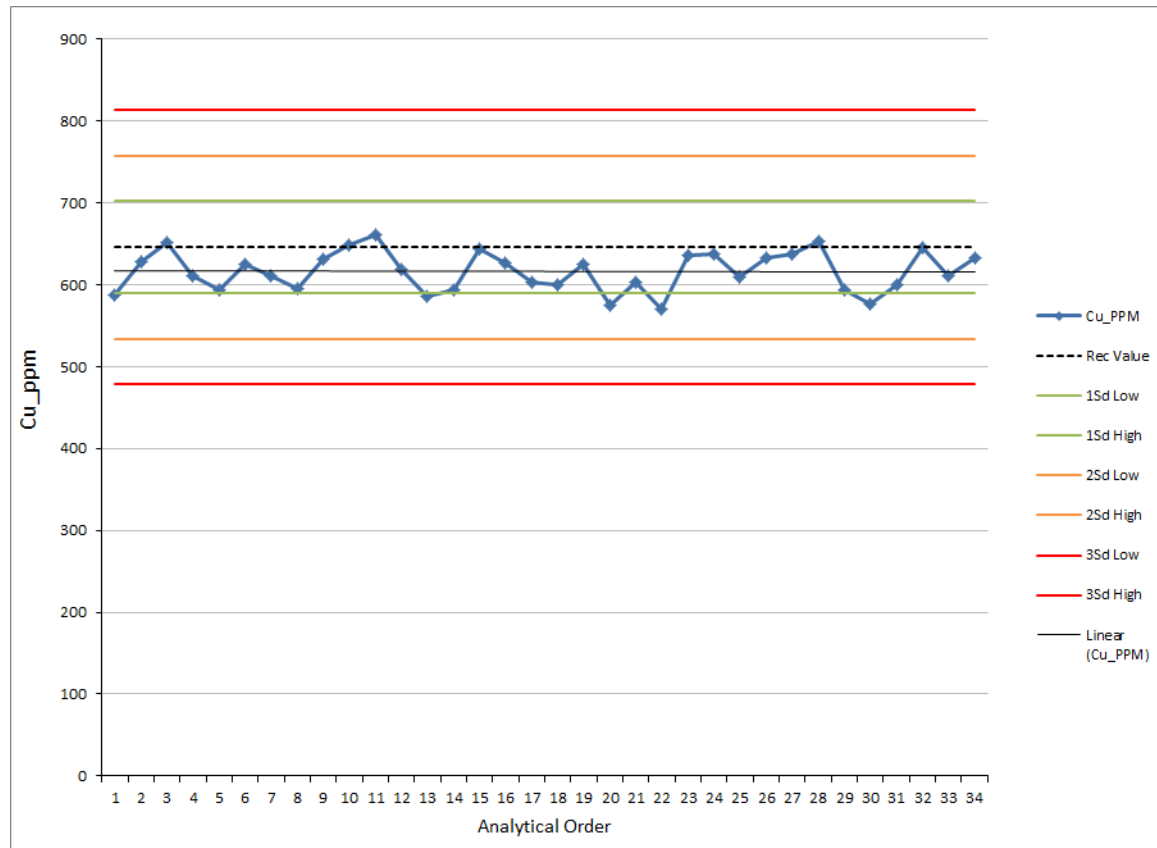


1 of 225 > 3 SD
= ~0.4%, which is also
statistically expected



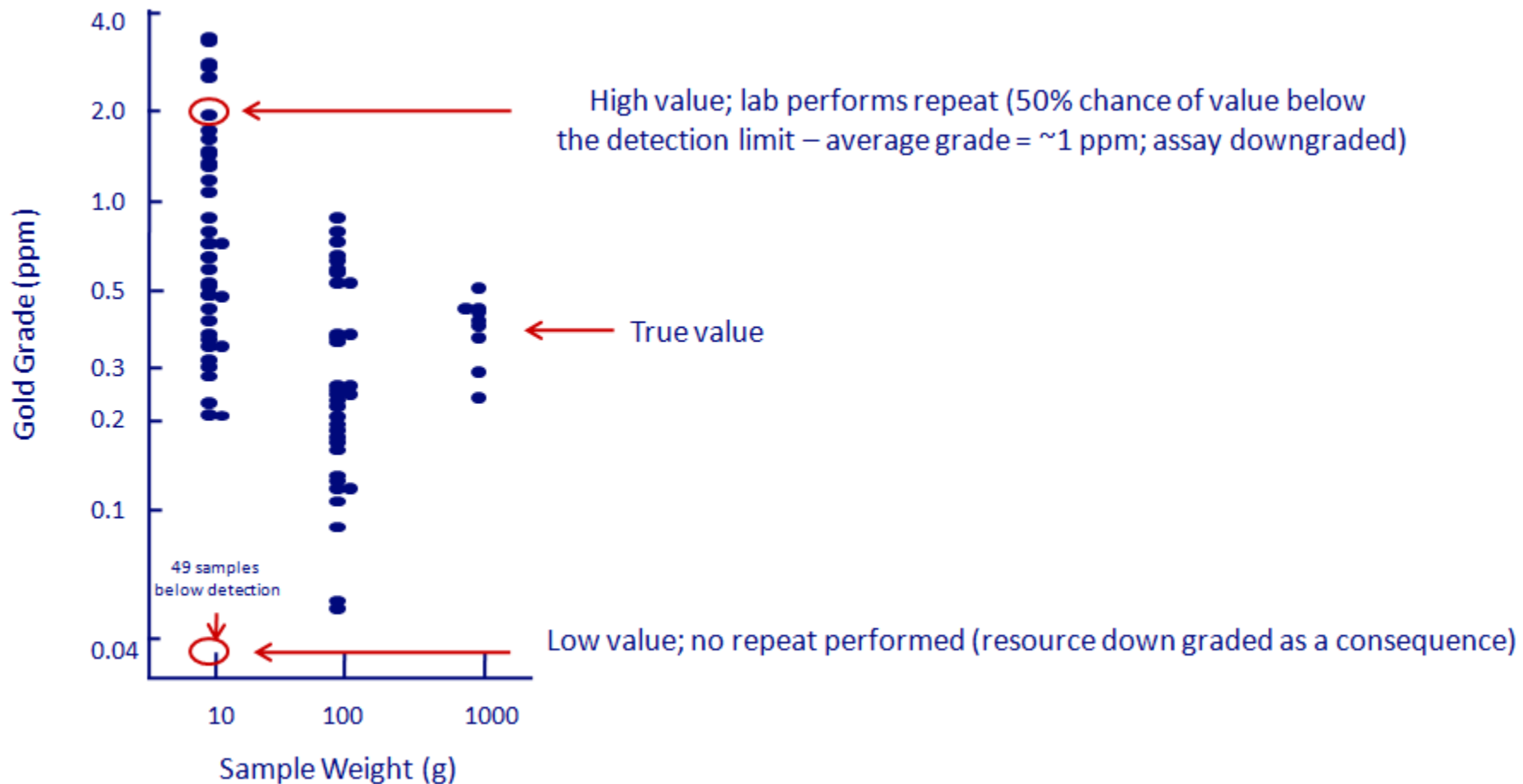
The Au results show the natural spread in data

What's an Acceptable Bias?



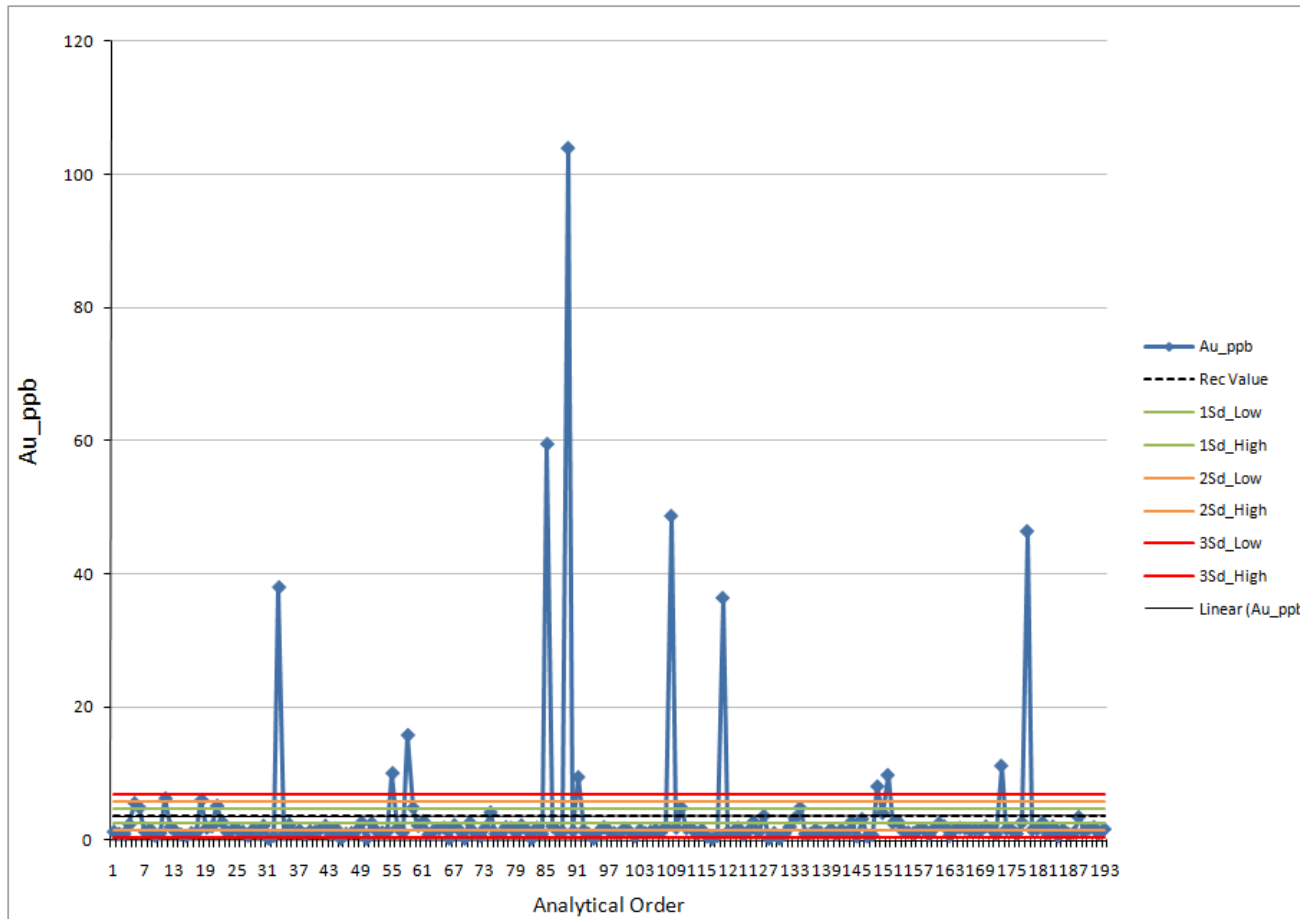
Bias is ~5 % - similar to the uncertainty of the certified mean

Induced Bias



A negative bias can be introduced into a data set by selectively re-assaying only high values and averaging the results.

Is Your Blank Contaminated?



- Locally sourced sand not analyzed to demonstrate homogeneity
 - Contains nuggetty Au (best anomalies!)

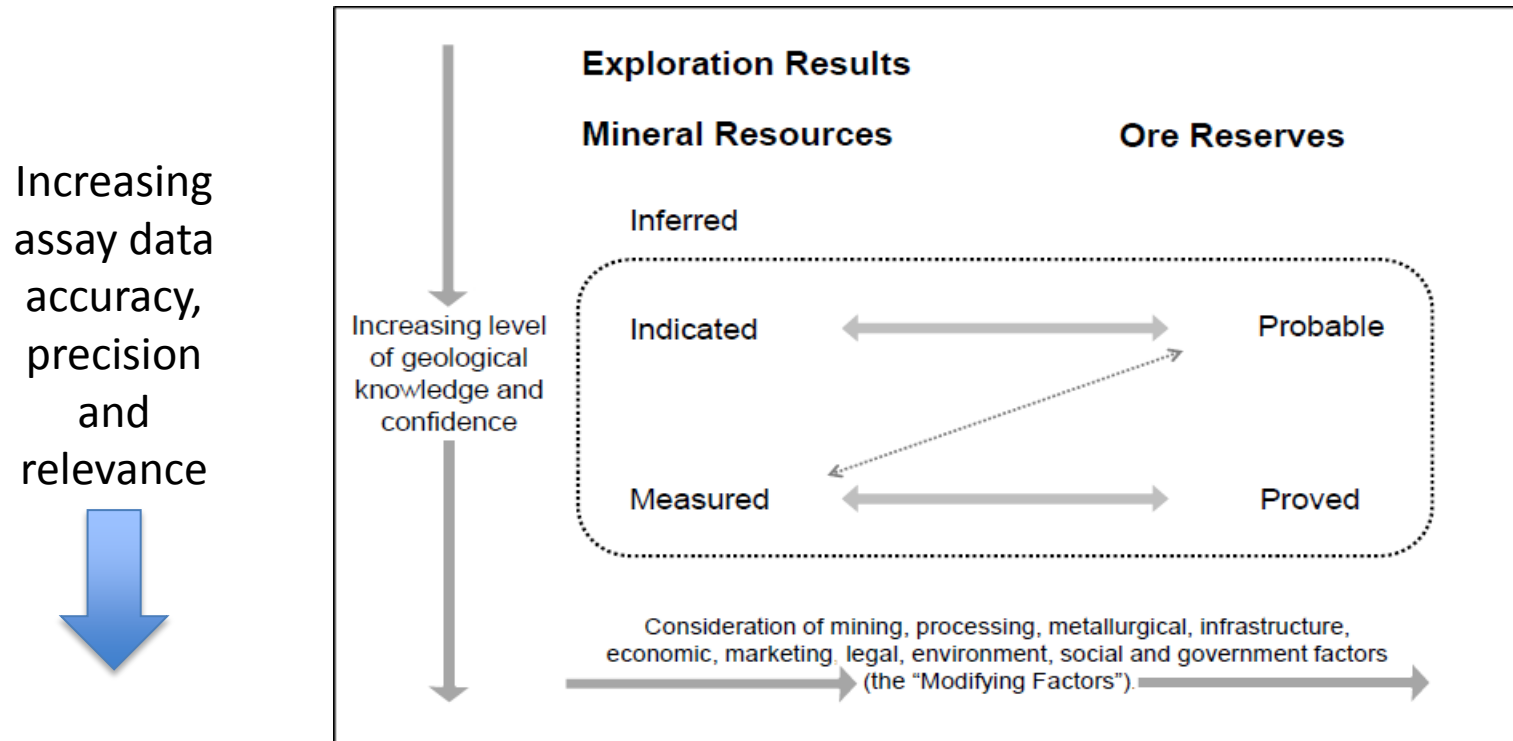
How do we Define Precision?

Table 1. List of Common Measures of Relative Error Calculated from Duplicate Pairs of Measurements and Used in a Variety of Geological Applications

Measurement	Conceptual Formula	Single Duplicate Pair Formula	Average Formula for Several Duplicate Pairs	Relationship with <i>CV</i>
Coefficient of Variation (<i>CV</i>)	$CV = \frac{\sigma}{\mu}$	$CV = \frac{2}{\sqrt{2}} \frac{ x_1 - x_2 }{(x_1 + x_2)}$	$\overline{CV} = \sqrt{\frac{1}{n} \sum_{i=1}^n \left(\frac{2}{\sqrt{2}} \frac{ x_{1i} - x_{2i} }{(x_{1i} + x_{2i})} \right)^2}$	<i>CV</i>
Relative Precision (<i>RP</i>)	$RP = \frac{2\sigma}{\mu}$	$RP = \frac{4}{\sqrt{2}} \frac{ x_1 - x_2 }{(x_1 + x_2)}$	$\overline{RP} = \sqrt{\frac{1}{n} \sum_{i=1}^n \left(\frac{4}{\sqrt{2}} \frac{ x_{1i} - x_{2i} }{(x_{1i} + x_{2i})} \right)^2}$	$2 \times CV$
Relative Variance (<i>RV</i>)	$RV = \frac{\sigma^2}{\mu^2}$	$RV = 2 \frac{(x_1 - x_2)^2}{(x_1 + x_2)^2}$	$\overline{RV} = \frac{1}{n} \sum_{i=1}^n \left(2 \frac{(x_{1i} - x_{2i})^2}{(x_{1i} + x_{2i})^2} \right)$	CV^2
Absolute Relative Difference (<i>ARD</i>)	$ARD = \frac{ x_1 - x_2 }{\mu}$	$ARD = 2 \frac{ x_1 - x_2 }{(x_1 + x_2)}$	$\overline{ARD} = \sqrt{\frac{1}{n} \sum_{i=1}^n \left(2 \frac{ x_{1i} - x_{2i} }{(x_{1i} + x_{2i})} \right)^2}$	$\sqrt{2} \times CV$
Half Absolute Relative Difference (<i>HARD</i>)	$HARD = \frac{1}{2} \frac{ x_1 - x_2 }{\mu}$	$HARD = \frac{ x_1 - x_2 }{(x_1 + x_2)}$	$\overline{HARD} = \sqrt{\frac{1}{n} \sum_{i=1}^n \left(\frac{ x_{1i} - x_{2i} }{(x_{1i} + x_{2i})} \right)^2}$	$\frac{\sqrt{2}}{2} \times CV$

Stanley, C.R. and Lawie, D., 2007, Average relative error in geochemical determinations: Clarification, calculation and a plea for consistency. *Exploration & Mining Geology*, vol. 16, pp. 267-275.

Quality Expectations



- Exploration: >90 % CRM pass rate; <10 % bias; CV<15 %
- Inferred resources: >95 % CRM pass rate; <5 % bias; CV<10 %
- Measured & indicated: >98 % CRM pass rate; <5 % bias; CV<5 %

Conclusions

- Errors are additive - it starts with sampling
- Set corporate data quality objectives
- Data should be fit for purpose
- Quality expectations should evolve with project
- Focus on the desired result (JORC), not the process (NI43-101)
- Don't forget the quality assurance
- As an industry we are imprecise on precision