

II INTERNATIONAL CONGRESS ON MINERAL CRUSHING, GRINDING, HPGR & CLASSIFICATION

# "New Drives for Better Geometallurgy Practices"

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Organized by



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# Agenda

- ✓ Introduction
- ✓ Multiple domains and variable hierarchy
- Non Linearity and uncertainty
- ✓ Modelling tools
- ✓ Case Studies
- ✓ Remarks

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#### **Geomet Environment**



# omminution 2019 Multiple Domain System – Attribute Hierarchy

Attribute Xi	Variance
Grade	S <sub>1</sub> <sup>2</sup>
Lithology	S <sub>2</sub> <sup>2</sup>
Alteration	S <sub>3</sub> <sup>2</sup>
Mineralogy	S <sub>4</sub> <sup>2</sup>
Blastability	S <sub>5</sub> <sup>2</sup>
Crushability	S <sub>6</sub> <sup>2</sup>
Copper Recovery	\$ <sub>7</sub> <sup>2</sup>
Acid Consumption	S <sub>8</sub> <sup>2</sup>

$$\begin{bmatrix} Sample \\ Amount \\ for X_1 \end{bmatrix} = f \left( \begin{bmatrix} Variability \\ of Attribute \\ X_1 \end{bmatrix}, \begin{bmatrix} Accepted \\ Confidence \\ Level \end{bmatrix} \right)$$
$$\begin{bmatrix} Sample \\ Amount \\ for X_2 \end{bmatrix} = f \left( \begin{bmatrix} Variability \\ of Attribute \\ X_2 \end{bmatrix}, \begin{bmatrix} Accepted \\ Confidence \\ Level \end{bmatrix} \right)$$

<u>COPPER GRADE</u> is unequivocally defined, but <u>COPPER</u> <u>RECOVERY</u> depends on process conditions

#### Omminution 2019 Non –Linear Responses



After Coward et al., 2009.

# omminution 2019 Most Geomet Processes are Non-Linear...



□ <u>Nil</u> Cl in electrolyte, <u>lower</u> EW-Cu grain cristallinity

□ *Low* Cl in electrolyte, *higher* EW-Cu grain cristallinity

□ *<u>High</u>* Cl in electrolyte, *Lower* EW-Cu grain cristallinity

#### Do you still think in Linear Processes?

### Jomminution 2019 Deterministic or Stochastic Planning?

Normally, the production plan is computed in deterministic mode. Asymmetry of the non-linear response, drives to biased estimate.

Random variability of input variables can be measured, then this can be included to estimate the production plan, using stochastic simulation, to find more realistic estimate of the production plan.



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#### **Some Current Grindability Tests**

Test	Mill Dia.	Top Size	Core	Database	
lest	m	mm		Y/N	
Bond Low-energy Impact	N/A	76.2	PQ/HQ	Y	
Media Competency	1.83	175		Y	
MacPherson Autogenous	0.46	32	NQ	γ	
JK Drop-weight	N/A	63	PQ/HQ	Y	
SMC Test <sup>®</sup>	N/A	31.5	Any	Y	
JK Rotary Breakage Test <sup>®</sup>	0.45	53	HQ	Y	
SAGDesign	0.49	38.1	NQ	Y	
SPI®	0.305	38.1	NQ	Y	
AG Pilot Plant	1.75	200		Y	
Lab-scale HPGR	0.25	12.7	BQ	Y	
SPT	N/A	19.1	BQ	Y	
HPGR Pilot Plant	0.90	50		Y	
Bond Rod Mill	0.305	12.7	Any	Y	
Bond Ball Mill	0.305	3.35	Any	Y	

after F.O. Verret et al., 2011

### omminution 2019 Grindability Test Performance

"Recommendations to reduce the error (SGI determination) as it ranges 25%..."

"To reduce error, the feed size distribution..."

"The grinding curve should be modelled using a variant of the Swebrec equation and..."

"A shorter, *lower cost/lower precision* version of the test has been recommended..."



Taken from: P. Amelunxen et al. / Minerals Engineering 55 (2014) 42–51

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#### **A Better Approach: Physical Quality Units**

Ore Attribute	Ox-03	Ox-04	Ox-01	Ox-02	SS-02	SS-03	Mix-01	Mix-02	SS-01
Alteration Ser.+Argíl.	18.4	14.3	14.1	13.9	12.1	12.0	0.6	0.5	0.7
Unit Weight., t/m <sup>3</sup>	2.55	2.53	2.58	2.53	2.62	2.56	2.66	2.67	2.62
UCS, MPa	42.98	50.62	57.21	58.79	56.43	71.67	72.33	91.21	91.57
TR, MPa	5.74	6.72	6.34	7.45	8.22	7.85	9.68	10.19	8.10
Young Index, GPa	29.8	32.2	33.0	35.1	39.3	38.1	42.3	44.1	46.6
RQD, %	80.16	80.51	73.40	75.47	94.38	91.26	92.93	92.93	94.84
FF, f/m	7.70	7.75	9.84	9.06	3.25	4.32	3.65	3.77	2.88
LRS, cm	163.15	166.37	153.29	155.92	190.68	186.12	189.13	189.13	191.70
GSI, %	43.78	45.88	42.96	44.91	55.37	53.73	53.94	54.72	54.82
RRD, %	3.09	3.43	3.35	3.41	3.97	3.94	3.86	3.89	3.92
BWI, kWh/t	7.8	10.8	10.5	11.3	12.5	12.9	15.8	16.1	16.4
SPI, min	38	65	54	67	97	102	130	126	140
Physical Quality Index	1	2			3		4		

The physical quality component naturally arise from the ranking of attributes.

# Jomminution 2019 Scale Up of Continuous Comminution Machines



- Most current methods consider only fracture phenomena
- Often severe simplification of the PBM approach is used
- Test scale is generally too small; geo structural features are missed

### omminution 2019 Simulate Instead of Estimate

![](_page_11_Picture_1.jpeg)

#### Dynamic Phenomenological Simulator

#### Block Model Linear

**Estimator** 

Tonnage Grades Density Lithology Mineralogy ... Recovery Acid Cons.

![](_page_11_Figure_6.jpeg)

Draw taken from Geometallurgy – Optimising the resource, R. Baumgartner, Geneva, 2012.

# omminution 2019 Modelling Approach for Planning

□ All characteristics of the orebody are <u>SPATIALLY VARIABLES</u>

□ Once extraction start all processes are <u>TIME DEPENDENT</u>

**NON LINEARITY** is everwhere

□ Metallurgical results are highly dependent on <u>PROCESS VARIABILITY</u>

![](_page_12_Picture_5.jpeg)

"Dynamic" modelling is appropriate and preferently in a phenomenological framework

# Planning with "Matrix Modelling"

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![](_page_13_Figure_1.jpeg)

# omminution 2019 Planning with "Matrix Modelling"

Example: Prediction of concave change in the primary crusher from geological data and process modelling.

![](_page_14_Figure_2.jpeg)

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Case Study 2: Budget Planning – Estimation versus Simulation

Draw taken from Geometallurgy – What, why and how, P. Lamberg, Levi, Finland, 2011

# Omminution 2019 **"The Operator Inequality at the Concentrator"**

$$P_{\text{Plan}} = mLR \ge \int_{0}^{303} Q_{P}(t) \rho_{P}(t) S_{P}(t) L(t) R_{F}(t) dt = P_{\text{Real}}$$

Planned Yr. Production, Block Model

![](_page_16_Picture_3.jpeg)

WON BY EXCESS IS LESS THAN LOST BY DEFFICIENCY IN ANY ASYMPTOTIC NON LINEAR PROCESS Real Yr. Production, Dynamic Process

This inequality is a permanent paradigm for the Concentrator plant operator, always trying to reconcilliate planned and operational results.

This is often due to an *incorrect planning* which ignores dynamic variability linked to process non-linearities and operational constrains.

# omminution 2019 Study Case: Simulate versus Estimate

Three mine programs: (i) <u>year</u>-basis, (ii) <u>monthly</u>-basis and (iii) <u>daily</u>-basis. The first two provided by the mine. The third one obtained by *Stochastic Simulation*.

![](_page_17_Figure_2.jpeg)

#### omminution 2019 Study Case: Simulate versus Estimate

![](_page_18_Figure_1.jpeg)

**Managing the Operational Variability...** 

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![](_page_19_Figure_1.jpeg)

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### Remarks

- □ It must be distinguished between *fundamental attributes* like Copper Grade and *responses* like Copper Recovery as the first one is an absolute independent quantity and the last one is a relative dependent quantity.
- □ Using linear estimation approach, such as <u>Kriging</u>, is not valid for non-linear and non-additive variables such as recovery of elements.
- □ It is hard to believe that current comminution tests and related interpretation methods may drive to accurate production estimates due to structural flaws.
- □ Most of the sizing equations are simplified versions of the PBM approach. <u>Generally speaking, none considers transport aspects</u>.
- □ For planning subjects we suggest to manage the "*Physical Quality*" concept properly included within the invariant transfer functions.

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![](_page_21_Picture_1.jpeg)

- For design and sizing of any industrial continuous comminution machine, we need to consider: (i) Comminution tests (mass-power-PSD), (ii) Rheological tests (yield stress, Bingham fluid), (iii) Validation continuous test to measure transport parameters and set scale up criteria.
- □ The experimental data must be fed to a phenomenological comminution model with two invariant transfer functions:
  - $\checkmark$  The reduced selection function.
  - $\checkmark$  The transport function.
- Next generation of comminution models (matrix models) will consider <u>explicitly</u>: Process, Operational and Maintenance aspects, among others.
- Variability not only intrinsic to the orebody but also related to actual operation in the mine and the plant, leads to <u>overestimate Production</u>. Better estimates of real results are obtained by <u>Random simulation</u>.