### Evidentiary Considerations for Integration of Biomarkers in Drug Development

# Statistical Considerations for Clinical Safety Biomarkers



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#### Potential biomarker panel for drug-induced pancreatic injury: Hypothetical example COU 1

Potential biomarkers:	Context of Use (COU 1):
<del>1. MiR-216a</del>	Claim: Qualified biomarkers to be used together
<del>2. MiR-375</del>	with conventional biomarkers, in early clinical drug development (in HV) to <b>support conclusions as</b>
3. Protein RA1609	to whether a drug is likely or unlikely to have
4. Protein RT2864	caused a mild injury response in the pancreas at the tested dose and duration.
5. Trypsinogen-1	<b>Research use:</b> To make decisions in real time on
6. Trypsinogen-2	individual or dose cohort based on changes in
7. Trypsinogen-3	<b>biomarker concentrations (from baseline)</b> , complementing the use of standard biomarkers

<u>Supportive studies</u>: Two prospective case/control studies in patients using medications that have potential to cause pancreatic injury:

- 1. Azathioprine in Crohn's disease patients
- 2. Mesalazine in ulcerative colitis patients with normal pancreas function
- Show greater diagnostic predictivity compared to amylase and lipase with a formal adjudication procedure and a predefined statistical evaluation



## Hypothetical example for drug-induced pancreatic injury COU 1 (cont.)

- Learn and confirm approach: ample learning completed at this stage
  - COU 1 clearly defined (support conclusions related to pancreatic injury response)
  - Objectives of confirmatory studies defined (greater diagnostic predictivity)
  - Biomarker panel chosen (though not clear from COU 1 how panel will be used, e.g., individual biomarkers or combination)
  - Measure of biomarker identified (e.g., dynamic change from baseline instead of single timepoint concentration)
- <u>Predefined statistical evaluation</u> of two prospective studies
  - Study results must support defined COU 1



### **Predefined statistical evaluation:** study results must support defined COU 1

- <u>Clear hypotheses regarding how biomarkers are to be</u> considered for use (relevant null and alternative):
  - E.g., using biomarkers + conventional markers relative to conventional markers alone will improve the sensitivity (or specificity) to identify patients treated (not treated) with medications known to potentially cause pancreatic injury
- Individual analysis to support each hypothesis
  - Lower bound 95% CI on difference > 0 (is 0 good enough?)
- But, <u>how to identify</u> patients as having potential injury response?
  - Signal in any 1 biomarker, signal in 2 of 3, signal in ALL, signal in a measure that combines and reduces 3 biomarker measures into 1 composite measure?
  - And, what is a "signal"? Predictive of injury? Predictive of exposure? Outside variation of HV? Is there a pseudo or true gold standard?



#### True gold standard vs "pseudo-gold standard"

- Gold standard (e.g., histopathology)
  - May be unavailable, too invasive, too expensive
  - If exists, new biomarker performance can be assessed through standard methods (e.g., ROC analysis) to show "comparability" to gold standard
- "Pseudo-gold standard" often inadequate (e.g., amylase/lipase in pancreatic injury lack specificity)
  - Comparing new biomarker using pseudo-gold standard as reference is unlikely to show improvement
  - Using <u>treatment (exposure) as a reference</u> possible to show improvement

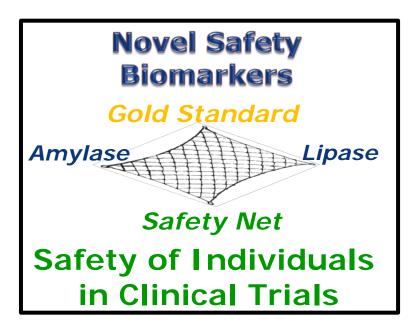
		Conventional markers only			
		Assessed as exposed	Assessed as NOT exposed	Total	
Biomarkers+	Assessed as exposed	А	В	A + B	
Conventional markers	Assessed as NOT exposed	С	D	C + D	
	Total	A + C	B + D	# controls	

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Specificity of conventional markers can be compared to that of biomarkers + conventional markers to show improvement (e.g., 95% CI LB > 0)

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What is the risk if the biomarker(s) lack predictive accuracy: Type I vs Type II error



**Type I error**: qualify biomarkers that do not predict toxicity

**Type II error**: reject biomarkers that do predict toxicity

Which is worse? Depends on intended use and current standard practice

- Intended use: to expand testing new drug when conventional biomarkers alone are considered inadequate (i.e., too risky)
   ensure biomarkers predict outcome (Type I error)
- Intended use: to conclude new drug is unsafe if biomarkers or conventional markers indicate it unsafe when conventional biomarkers alone are considered adequate
   ensure identify potential injury (Type II error)



### Predefined statistical evaluation: agreement of analytical plan

- Pre-defined statistical analysis plan to address:
  - How to combine data from multiple studies (pooling, meta-analysis)
  - How to handle missing data (ignore/remove, LOCF, imputation)
  - What are important **sensitivity analyses**?



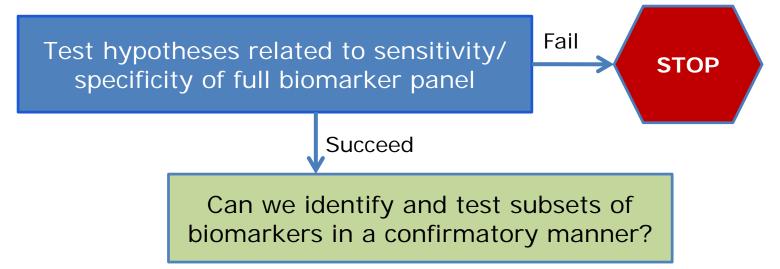
### Additional considerations: adaptive strategy to continue learning while confirming?

Interim Analysis	Timing of Interim Analysis	Purpose of Interim Analysis	Example Rule
1 (IA 1)	After completion of ~ first 25% of all study data (first ~25% from each prospective studies)	<ul> <li>Assess initial performance to with respect to sensitivity/ specificity hypotheses</li> <li>Potential to modify biomarker rules to identify "signal"</li> <li>Potential to increase sample size</li> </ul>	<ul> <li>If observed specificity         &lt; 80%, modify biomarker         rules. Exclude data from IA         1 in final analysis, increase         overall sample size so final         analysis is fully powered</li> <li>If observe specificity ≥ 80%         continue to final analysis</li> </ul>
2 (perform only if modify rules at IA 1)	After completion of ~ second 25% of all study data (second ~25% from each prospective studies)	<ul> <li>Assess initial performance of modified rules with respect to sensitivity/specificity hypotheses</li> <li>Potential to stop prospective studies for futility</li> </ul>	<ul> <li>If observed specificity &lt; 80%, stop studies for futility</li> <li>If observe specificity ≥ 80% continue to final analysis</li> </ul>

#### What is impact on Type I/Type II error? Simulations are useful



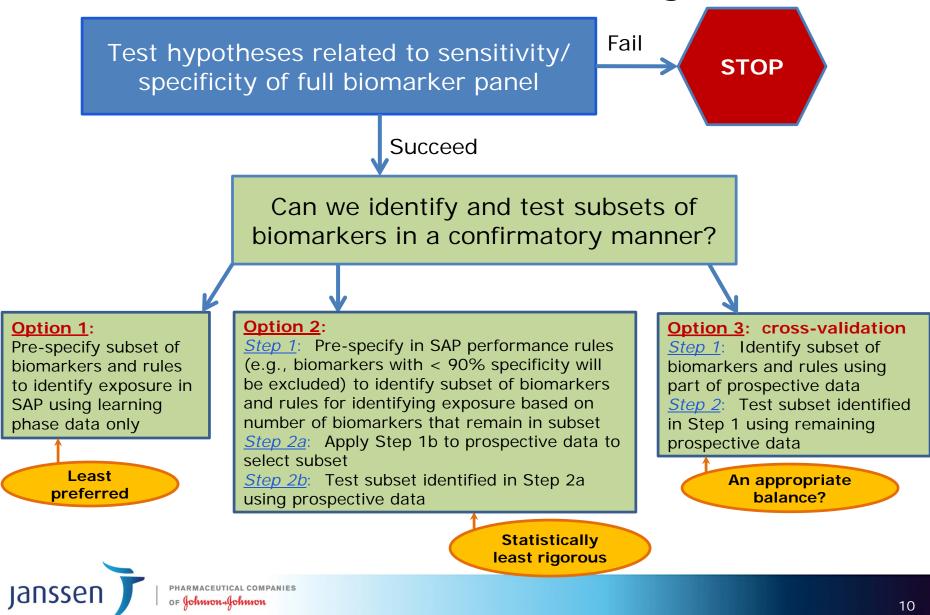
### Additional considerations: can we explore biomarker subsets while confirming?



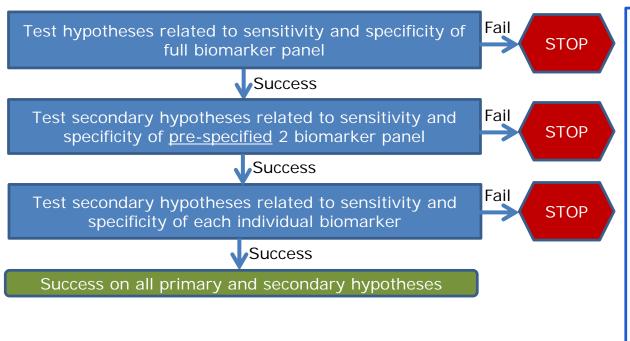


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### Additional considerations: can we explore biomarker subsets while confirming?



### Additional considerations: Option 1 to explore biomarker subsets



May be difficult to pre-specify and identify subsets when the number of biomarkers in the panel is > 3

A hierarchical testing strategy was proposed to protect the overall Type I error at  $\leq 2.5\%$  (1-sided)

- Both sensitivity and specificity tested at each level, success on both must be met to proceed to the next level
- Within final level of the hierarchy, the sensitivity and specificity of the 3 individual BmXs can be tested using appropriate multiplicity adjustment (e.g., Hochberg)



#### Potential biomarker panel for drug-induced pancreatic injury: Hypothetical example COU 2

Potential biomarkers:	Context of Use (COU 2):
1. Protein RA1609	<b><u>Claim</u></b> : A composite measure (CM) of the qualified biomarkers to be used together with conventional
2. Protein RT2864	biomarkers, in normal healthy volunteer trials
3. Trypsinogen-3	supporting early clinical drug development
	<b>Research use:</b> to make decisions in real time on
	dose cohort using group average of CM, based on
	changes in biomarker concentrations (from

biomarkers <u>Supportive Data</u>: Learning phase data to support objectives for COU 1 One study in healthy subjects at 2 visits, and one study in patients with

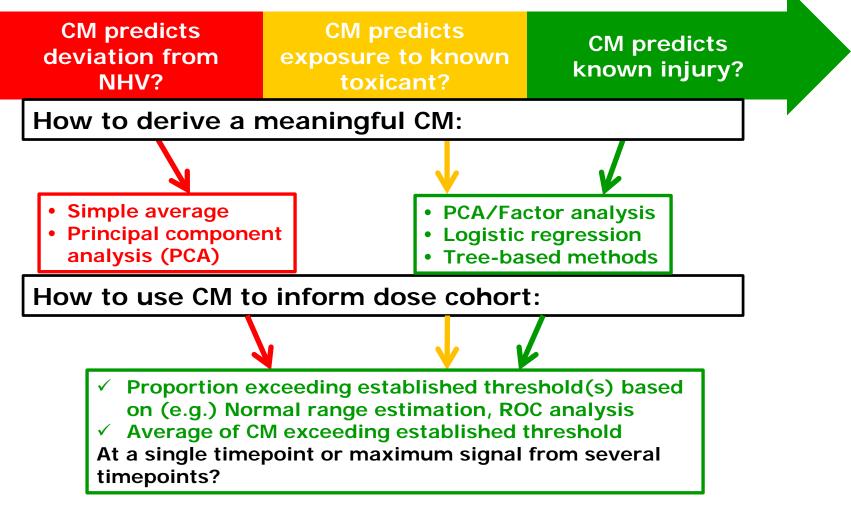
baseline), complementing the use of standard

known pancreatic injury

✓ Characterize expected variability of CM in NHV and show association of CM with known injury



### Hypothetical example for drug-induced pancreatic injury COU 2 (cont.)

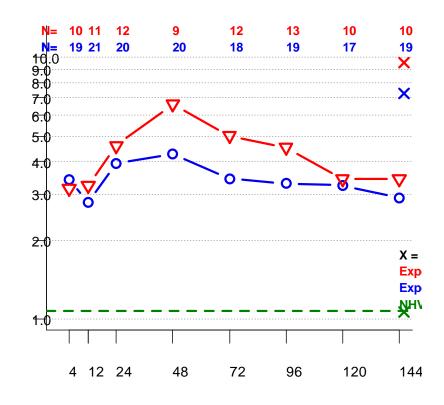


#### • What are the limitations of the learning data?



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#### Some potential limitations of learning data



Hour post-baseline

Individual Patient CM = GM of 3 BmX FC from BL GM CM = GM of Individual Patient CMs

- May only confidently use to predict deviation from NHV
- Multiple timepoints for exposed patients, limited timepoints for NHV
- Signal much larger using maximum across all timepoints
- Association ≠ Causation
- How can we derive thresholds?
  - Bootstrap, but only for single timepoint
  - Modeling and simulation, with assumptions

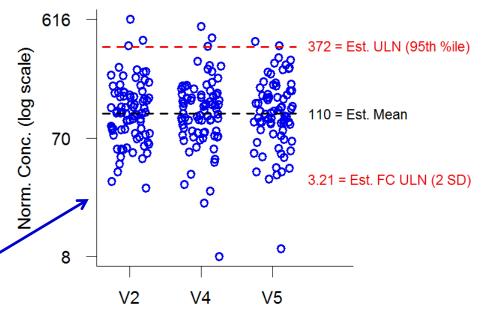


Geometric Mean

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#### Other relevant statistical considerations before COU 1/COU 2

- What is the right biomarker measure?
  - Raw concentrations, normalized concentrations, change from baseline (absolute or fold-change)
- How to estimate normal ranges (i.e., in NHV)?
  - "robust" (Horne and Pesce) method, non-parametric bootstrap, assumptions of normality (can transform)





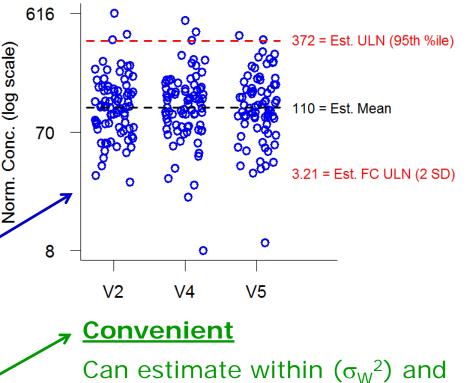
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#### Potential effects of covariates



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Can estimate within  $(\sigma_W^2)$  and between  $(\sigma_B^2)$  subject variability

If  $\sigma_W^2 << \sigma_B^2 \Leftrightarrow$  change

If  $\sigma_B^2 >> \sigma_W^2 \Leftrightarrow absolute$ measure

#### Other relevant statistical considerations before COU 1/COU 2 (cont.)

- Selection of biomarkers
  - <u>Many statistical methods</u>: regression (traditional, ridge, LASSO), classification/ROC, tree-based methods
    - Multiplicity concerns can be mitigated using false discovery rate methods and cross-validation
  - Selecting a few among potentially many typically goes beyond statistics

Biomarker	Performance in Learning Studies	Biological Interpretation	Assay Availability and Confidence – e.g., LLOQ/ Analyte Stablility/ No Special Buffer needs	Translatability	Cost
1					
2					

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### **Concluding remarks**

- Defining universal evidentiary standards for safety biomarker qualification is difficult
  - Significant diversity in potential context of use
- Appropriate evidentiary standards rely on core statistical principles
  - Some may mimic traditional evidentiary standards associated with drug development (Clear hypotheses, analyses, multiplicity, missing data, ...)
  - Some may not (Settings in safety qualification where Type II error may be important, integrating more than one study for final analysis, ...)
- Key beyond statistics: cooperative efforts (consortium), regulatory interactions, patience



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