QUANTITATIVE DECISION-MAKING IN DRUG DEVELOPMENT

EFSPI Statistical Leaders meeting

July 4th, 2017

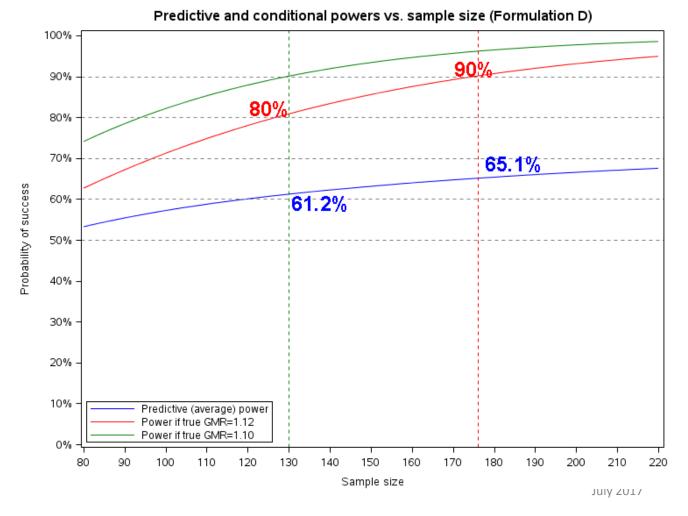
Maylis Coste & Sylvain Nicolas

Simple cases, at study level

Bayesian PoS for a PK study where variability/uncertainty is high Assumption:

True ratio assumed to be around 1,12 (with uncertainty as 90%CI = 1,12 [0,93 - 1,35]) SDw assumed around 0,35

Predictive power even lower when using more complex covariance matrices



Objectives

Assess our current practices

Discuss our expectations about the contribution of the statisticians, as well as the Statistical Leaders, into Decision Making

Define how to achieve this together

Decision-making in drug development

At a given time point, making an optimal choice between several alternatives based on the available information and preferences of the decision maker

Study level

- Choice of the dose
- Population, design (sample size, control arms, duration)
- Stop/continue at interim analyses

Development level

- Strategy: indication, population, Number of studies, timing of the studies
- Go/No Go at strategic milestones
- Due diligences
- Global project value assessment

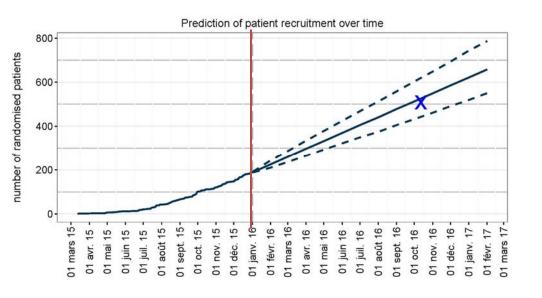
Portfolio level

- Go/No Go and selection of the projects
- Resource allocation

Bayesian modeling for end of study prediction

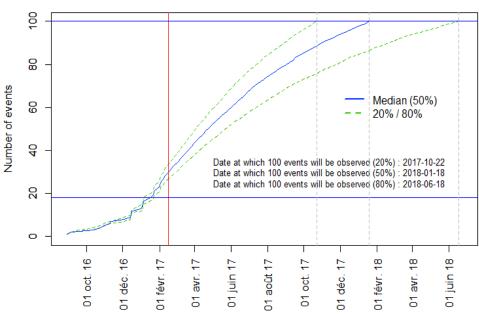
Recruitment prediction

Date of end of recruitment estimated with a 95%CI



Target nb of events prediction



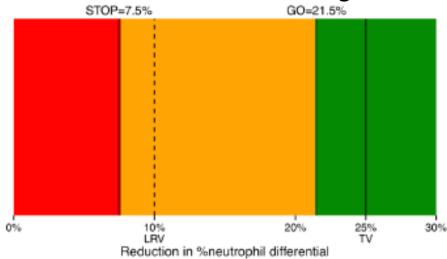


AstraZeneca's decision-making framework Software developed by Cytel

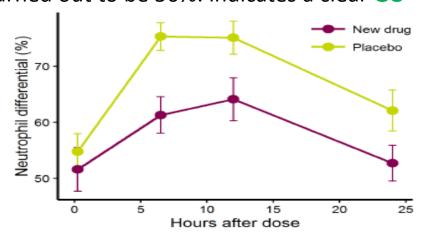
Ex.: Go/No Go criteria for neutrophil differential used as a biomarker for CPOD

1) Decision framework (Go/No Go/Think)

LRV: Lower Reference value – TV: Target value



3) Results: the observed level of reduction turned out to be 56%; indicates a clear **GO**



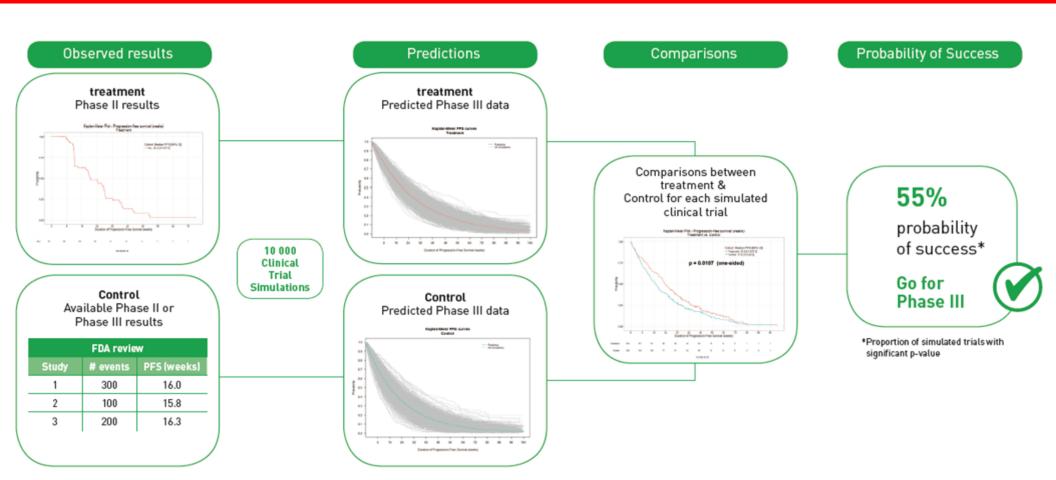
2) Operating characteristics

	Probability of different decisions under different true effects				
True effect (reduction)	Go	Indecisive	Stop	Go or Indecisive	Stop or Indecisive
TV (25%)	60%	30%	10%	90%	40%
LRV (10%)	20%	38%	42%	58%	80%
Placebo (0%)	6%	25%	69%	31%	94%

Systematic approach requested by the governance boards in AstraZeneca Early Clinical Development

Source: Taib, Z. and Jauhiainen, A. (2016). 5th Early Phase Adaptive Trials Workshop, Politecnico di Torino, Sep 29 - Oct 1st, 2016.

Predictive Probability of Success of Phase 3 based on Phase 2 results and historical data (fictive case-study)

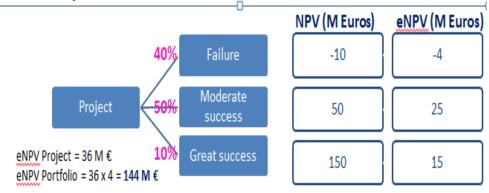


→ PPoS helps decision-making at strategic milestones

Comparison of Portfolios Basic fictive example

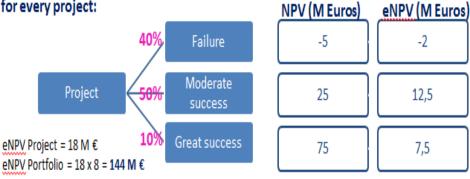
Portfolio strategies (with/without partner)

"Internal portfolio": 4 projects owned internally, with for every project:



"Partnered portfolio": 8 projects with 50% of the costs and revenues shared, with

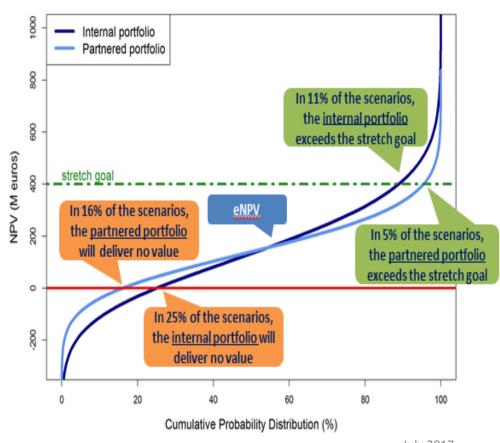
for every project:



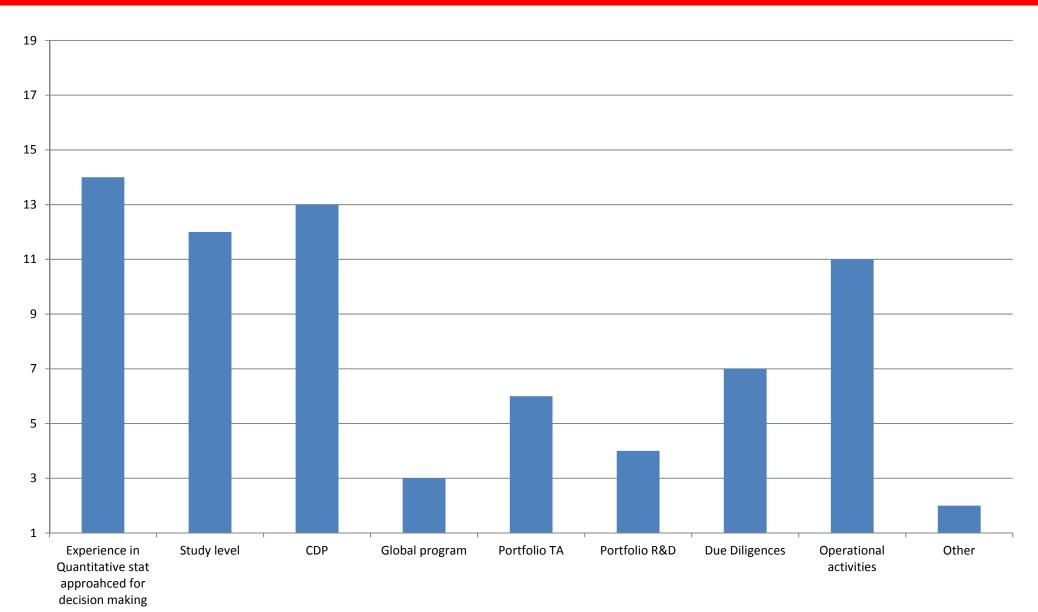
NPV: Net Present Value

eNPV: Expected Net Present Value

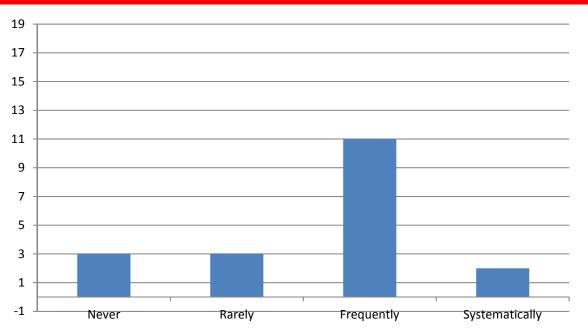
Risk and value profiles



Survey: Use of Quantitative decision making approaches by context

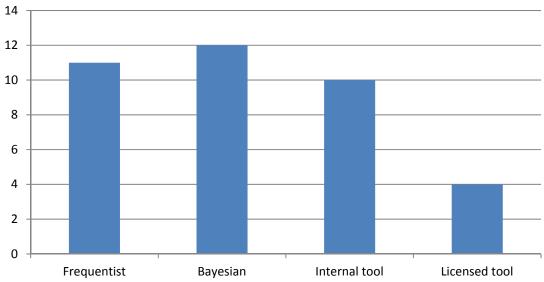


Survey: Frequency, Approaches and Applications



Frequency

Approaches and Applications



Points for discussion

- Confirm the strategic importance for Statistical teams to develop Quantitative supports for decision making
- Promote the position of Statistical teams/leaders to generate
 Indicators and manage their interpretation and impact
- Available for sharing methods and experiences
- Support the proposal of an EFSPI SIG to share
 - Change in mindset/culture and maybe organization
 - Decision criteria
 - Issues, methodologies, statistical methods/tools, interpretation & impacts
 - Examples of applications and stakeholders
 - (Fictive) Case studies and training(s)
 - Leadership : tbd
 - Members: Sanofi, Servier,tbd

Bibliography

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Incorporating budget considerations

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