

# **IMPACT:**

## **Imaging Metrics for Precision Analysis of Clinical Trial results**

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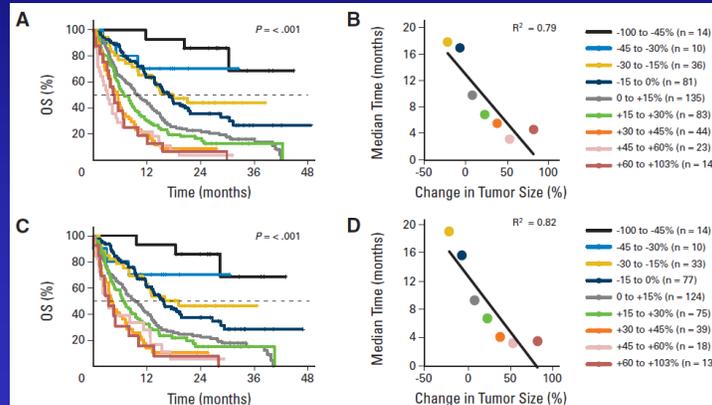
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## **Problem statement**

- **Oncology drug development is inefficient**
  - 62.5% of phase III trials are negative
- **Therapeutic progress has inherently made drug development more difficult**
  - Greater use of randomized phase II trials
  - However, trials continue to study traditional endpoints (ORR, PFS)
- **A comprehensive effort to develop new trial endpoints is needed**

## Background (1)

- It has been shown that a greater magnitude of response is associated with a better prognosis for an individual patient

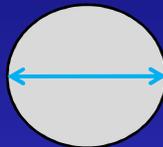


Jain et al, JCO, 2012

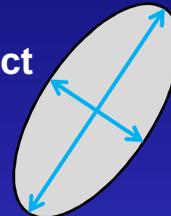
## Background (2)

- Advanced imaging of whole tumor volume can better characterize the entire biology of tumor growth and response

Diameter  
(RECIST)  
1D



Cross-product  
(WHO)  
2D



Volume  
3D



## **Background (3)**

- **Some have suggested that different analytical tools will not improve clinical trial analysis:**
  - Kaiser, CCR, 2012 compared PFS to growth modeling by re-sampling phase II trials
  - Used 5 large Genentech studies published in 2001-2005
  - Concluded that PFS is the best endpoint for phase III trial prediction
- **We prefer to study source imaging data to ensure highest quality data input**

## **Hypothesis**

1. **Quantitative analysis of tumor response as a continuous variable will improve the ability of randomized phase II trials to accurately predict phase III results**
2. **Detailed assessment of the entire tumor burden using volumetric CT will improve efficiency and accuracy of phase II trial analysis**

## **Aims**

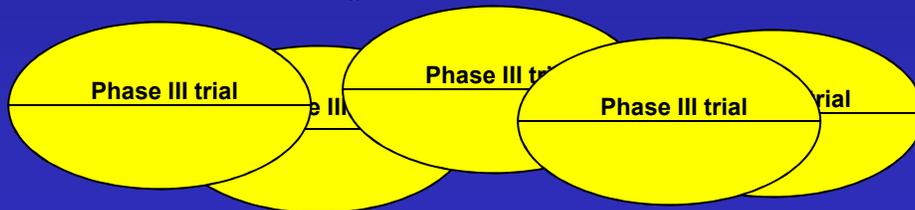
- 1. Assess feasibility of collection and analysis of images from completed phase III trials to:  
(A) simulate of phase II trial results and  
(B) develop quantitative metrics for improved prediction of phase III trial results**

## **Aims**

- 1. Assess feasibility of collection and analysis of images from completed phase III trials to:  
(A) simulate of phase II trial results and  
(B) develop quantitative metrics for improved prediction of phase III trial results**
- 2. Assess which quantitative metrics most accurately and reliably predict phase III results across different trials**
- 3. Quantify the added value of volumetric tumor measurement as compared to conventional measurement only**

## Approach (1)

- 1) Collection of existing trial data
  - Focus on trials with greater than ~150 patients per arm
  - Measurable carcinomas: NSCLC, RCC, CRC
  - Collect DICOM imaging from core labs holding scans for pharma
  - IRB has approved receipt of these de-identified images at Columbia

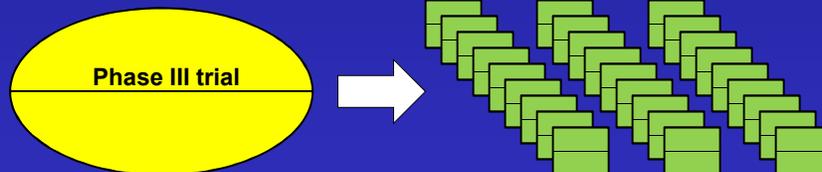


## Approach (2)

- 2) Generate semi-automated tumor measurements
  - DICOM images will be studied at a lab experienced with volumetry (e.g. Schwartz lab, Columbia University)
  - Computer generated tumor contours will be corrected as needed by an experienced technician
  - Measurements in 1D, 2D, 3D will be calculated for all lesions  $\geq 1$ cm (up to 10 lesions) at each time point

## Approach (3)

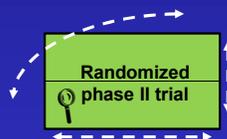
- 3) Develop simulated randomized phase II trials based upon existing trial data
  - Begin with measurement data from large completed clinical trials
  - Taking subsets of patients, will simulate multiple phase II trials of N patients
  - Simulation will incorporate measurement variability



Moskowitz et al, EJC, 2009

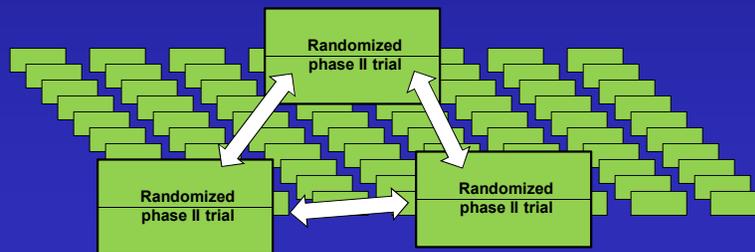
## Approach (4)

- 4) Comprehensively study each simulated randomized phase II trial with multiple metrics
  - Entire spectrum of measurement data will be studied, not just “best response”
  - Eventually will include statistical modeling of tumor growth & regression



## Approach (5)

- 5) Compare multiple simulations of the same trial to assess the reliability of each metric
  - The variance of each metric will be calculated across 1000 simulations
  - Change in variance with change in N will be studied for each metric



## Approach (6)

- 6) Correlate each trial metric with the hazard ratio (HR) from the parent phase III trials in multiple ways:
  - Pearson and rank correlation
  - Linear regression
  - ROC curves on various dichotomized versions of the HR
  - Sensitivity/specificity/predictive values on various dichotomized versions of the metric and HR

## **Value statement**

- 1. New metrics could provide greater clarity for go/no-go decisions regarding phase III drug development**
- 2. More efficient phase II trials will allow earlier results and more innovative studies (dose finding, subset analyses)**
- 3. The metrics from the proposed analysis could then be applied to other settings like biomarker development and prognostication**

## **Key strengths**

- 1. Collaboration between multiple academic and pharma parties, with FDA representation**
- 2. Comprehensive analysis of source image data**
- 3. No bias towards a specific trial analytic**
- 4. FNIH supported effort in pre-competitive space**

## **Progress**

- 1. Multiple pharmaceutical sponsors have interest in releasing data for analysis:**
  - Sanofi, GSK, Genentech, Celgene**
- 2. Positive response to date from ICL's for making imaging data available**
- 3. Seed funding for year 1 expected from FNIH, additional funding is needed**