FDA-AACR Public Workshop on OPTIMIZING DOSAGES FOR ONCOLOGY DRUG PRODUCTS: QUANTITATIVE APPROACHES TO SELECT DOSAGES FOR CLINICAL TRIALS

February 15, 2024, 8 a.m. – 5 p.m. February 16, 2024, 8 a.m. – 1 p.m. Grand Hyatt Washington | Washington, D.C.







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Qi Liu, PhD U.S. Food and Drug Administration

OPTIMIZING DOSAGES FOR ONCOLOGY DRUG PRODUCTS: QUANTITATIVE APPROACHES TO SELECT DOSAGES FOR CLINICAL TRIALS

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SECTION 3: SELECTING DOSAGES FOR REGISTRATIONAL TRIALS









Quantitative Approaches to Select Dosages for Clinical Trials

Considering the Totality of Efficacy and Safety Data to Aide Registrational Trial Designs

Stacy S. Shord, PharmD, BCOP, FCCP Deputy Division Director Division of Cancer Pharmacology II Office of Clinical Pharmacology OTS/CDER/FDA

February 16, 2024



Disclaimer

Opinions presented are those of the speaker and should not be construed to represent FDA's views or policies.



- Short observation period for DLTs
- Emphasis on DLTs, but not other safety or data

*DLT = Dose-limiting toxicity *MTD = Maximum tolerated dose

Possible Dosage Selection Strategy





Select Dosage for Further Exploration Evaluate Several Dosages

Holistic Approach



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How do we systematically evaluate all available nonclinical and clinical data?



Adapted from Dr. Youwei Bi update on MIDD program

PK = pharmacokinetics, PD = pharmacodynamics, PBPK = physiologically based pharmacokinetics, CiPA = comprehensive in vitro proarrhythmia assay

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Model-Informed Drug Development

- Leverage a thorough understanding of the drug, a disease, and how a drug affects the human body, as well as how the body responds to the drug.
- Quantify information by developing mathematical models based on full use of all available data, from sources such as in vitro, nonclinical and clinical studies
- Apply this knowledge to address issues pertaining to drug development or clinical use.

Adapted from CDER Conversation with Dr. Madabushi

MIDD Can Facilitate Drug Development



- Predict concentrations at different moments in time, including doses and times not yet studies
- Test effects of covariates to identify differences in exposure in specific subpopulations
- Characterize dose- and exposure-response relationships at any stage of development
- Facilitate a thorough understanding of the therapeutic index
- Leverage published data to help understand drug class effects and inform trial design

Considering the Totality of Efficacy and Safety FDA Data to Aide Registrational Trial Designs

(1)

Using Modeling-Based Approaches to Understand Doseand Exposure-Response Relationships for Activity Dr. Jin Y. Jin

(2)

Using Pharmacokinetic-Pharmacodynamic Modeling and Simulation to Understand Dose and Exposure-Response Relationships for Adverse Reactions Dr. Scott Van Wart



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Using Modeling-Based Approaches to Understand Dose- and Exposure-Response Relationships for Activity

Jin Y. Jin, Ph.D. Executive Director and Senior Fellow Global Head of Modeling & Simulation Clinical Pharmacology Genentech Inc. South San Francisco, CA

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Jin Y. Jin

I have the following relevant financial relationships to disclose: Employee of: Genentech, A Member of the Roche Group Stockholder in: F. Hoffmann-La Roche

The views expressed in this presentation are my own and not reflective of Roche/Genentech









- Challenges and Opportunities for Dose Optimization in Oncology
- Modeling-Based Approaches for Dose Selection (Registrational Trial)
 - **PERJETA** mAb, solid tumor (*population PK & exposure target*)
 - Lunsumi
 [TDB, heme (QSP, RO-based exposure-response)
 - Ipatasertib SM, solid tumor (biomarker PK/PD, exposure-response, clinical utility index)

Summary

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PD – pharmacodynamics, PK – pharmacokinetics, QSP – quantitative system pharmacology, RO – receptor occupancy, SM – small molecule, TDB – T-cell dependent bispecifics, TI – therapeutic window



Overview Challenges and Opportunities for Dose Optimization in Oncology



- Therapeutic window
- Translation from preclinical-to-clinical and early-to-late clinical
- Phase 1 in patients both challenge and opportunity
- Biomarkers (target, pathway, disease; circulating/biopsy; imaging/digital)
- PK confounding
- Immunogenicity (ADA)
- Confounding factors in patients
- Confounding of dose reduction, hold, missing
- Dose by body weight or flat dosing, administration route
- Combinations (PK DDI, efficacy, safety)
- Special population (pediatric, geriatric, renal/liver dysfunction, race/ethnicity)





ADA – anti-drug antibody, ADC – antibody-drug conjugate, DDI – drug-drug interaction, LM – large molecule, PK – pharmacokinetics, SM – small molecule, TDB – T-cell dependent bispecifics, TW – therapeutic window ©2024, Genentech CONFIDENTIAL FDA-AACR Public Workshop On



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Case Example #1

Pertuzumab Dose Selection Based on Clinical PK and M&S



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PERJETA

pertuzumab for injection



Dose regimen Trial simulation based on population PK model

- □ Trough concentration >20 µg/mL target
 - Loading dose for Cycle 1 C_{trough} >20 µg/mL
- Dose to be in linear range of PK: 2~15 mg/kg
 - Saturate receptor-mediated clearance
 - Reduce clearance variability
- Low incidence of immunogenicity
- □ No effect of body weight on PK
- □ 840 mg loading dose followed by 420 mg Q3W maintain trough concentrations above the 20 µg/mL target in >90% of patients in all cycles based on M&S

US Label

12.3 Pharmacokinetics

Pertuzumab demonstrated linear pharmacokinetics at a dose range of 2-25 mg/kg. Based on a population PK analysis that included 481 patients, the median clearance (CL) of pertuzumab was 0.24 L/day and the median half-life was 18 days. With an initial dose of 840 mg followed by a maintenance dose of 420 mg every three weeks thereafter, the steady-state concentration of pertuzumab was reached after the first maintenance dose.

The population PK analysis suggested no PK differences based on age, gender, and ethnicity (Japanese vs. non-Japanese). Baseline serum albumin level and lean body weight as covariates only exerted a minor influence on PK parameters. Therefore, no dose adjustments based on body weight or baseline albumin level are needed.

Garg A et al. Cancer Chemother Pharmacol. 74(4):819-29 (2014)

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Case Example #2 Mosunetuzumab Dose Selection Based on Mechanism-Based Model-Informed Strategy









Mosunetuzumab is a CD20/CD3 Bispecific Antibody for B-cell Malignancies:

- PK properties enable q3w dosing
- Does not require *ex-vivo* T-cell manipulation
- Off the shelf, readily available treatment

Mechanism of Action:

- Redirects T cells to engage and eliminate malignant B cells
- Potent tumor killing as single agent

Development Challenges:

- On-target acute toxicity for T-cell directing therapies
 - Cytokine release syndrome (CRS) occurs acutely following first doses and dissipates with time
- Challenges in dose finding
 - Complex target engagement
 - No simple PK target based on preclinical data
 - Phase I dose-response relationship was confounded by patients' prior lines of therapies and on-board residual rituximab

Budde LE et al. J Clin Oncol. 40(5):481-491 (2022) Budde LE et al. Lancet Oncol. 23(8):1055-1065 (2022)



Case Example #2 Mosunetuzumab Step-Up Dosing Strategy Supported by QSP Modeling







Hosseini I, et al. NPJ Systems Biology and Applications. 6: 28 (2020); Susilo ME, et al. Clinical and Translational Science. 16: 1134-48 (2023)

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Case Example #2 Characterizing the Driver (RO) for Mosunetuzumab Clinical Response





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Li C-C, et al. Blood. 134 Suppl 1: 1285 (2019); Modified from: Chi-Chung Li, ACoP (2019), Brendan Bender, ACoP (2020), Chi-Chung Li, Certara Webinar (2021)

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Case Example #2 Characterizing the Driver (RO) for Mosunetuzumab Clinical Response

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Li C-C, et al. Blood. 134 Suppl 1: 1285 (2019); Modified from: Chi-Chung Li, ACoP (2019), Brendan Bender, ACoP (2020), Chi-Chung Li, Certara Webinar (2021)

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Model-Based Dosing Strategy Expands the Therapeutic Window and Accelerates Mosunetuzumab Development





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| Table 1. Rec | ommend | ed LUNSUMIO I | US L Dose and Schedule (21-Day Treatment Cycles) | | |
|------------------|--------|---------------------|--|--|--|
| Day of Treatment | | Dose of LUNSUMIO | Rate of Infusion | | |
| Cycle 1 | Day 1 | 1 mg | Administer over a minimum of 4 hours. | | |
| | Day 8 | 2 mg | | | |
| | Day 15 | 60 mg | | | |
| Cycle 2 | Day 1 | 60 mg | Administer over 2 hours if infusions from Cycle 1 were | | |
| Cycles 3+ | Day 1 | 30 mg | well-tolerated. | | |

Li C-C, et al. Blood. 134 Suppl 1: 1285 (2019); Modified from: Chi-Chung Li, ACoP (2019), Brendan Bender, ACoP (2020), Chi-Chung Li, Certara Webinar (2021)

Ipatasertib Phase 2 Dose Selection: Biomarker PK/PD The PI3K/AKT pathway is central for cancer cell growth and survival Ipatasertib (GDC-0068) is a potent, oral, ATP-competitive AKT inhibitor

Optimal biological dose selected for Phase 2 based on target specific biomarker response and PK/PD



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Case Example #3

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 Median Observed

ADMINISTRATION

400 600ma

200

 ∞ Ö

Case Example #3

Ipatasertib Phase 3 Dose Selection: Account for Confounding of Dose Reduction









Zhu R et al, CPT Pharmacometrics Syst Pharmacol. 8:240–248 (2019)

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Case Example #3 Ipatasertib Phase 3 Dose Selection: Exposure-Response Analyses

Exposure-Efficacy: radiographic PFS



Dose-rPFS projections from Cox regression model of exposure-rPFS coupled with dose intensity model

Exposure-Safety: Gr2+ Diarrhea

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Dose-safety projections from logistic regression model of exposuresafety coupled with dose intensity model (Gr2+ diarrhea*)

*Similar analyses conducted for: Gr3+ diarrhea, Gr2+ rash, Gr3+ rash

Zhu R et al, CPT Pharmacometrics Syst Pharmacol. 8:240–248 (2019)

Case Example #3 Ipatasertib Phase 3 Dose Selection: Clinical Utility Index

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Benefit-risk analysis via exposure-response and clinical utility index (CUI) approaches indicated that Ipatasertib 400 mg QD has the highest probability of achieving the minimal Product Profile (PP) with best benefit/risk balance, which was thus selected for Phase 3



Zhu R et al, CPT Pharmacometrics Syst Pharmacol. 8:240–248 (2019)

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Summary





- Dose optimization strategy should be seamlessly integrated into clinical development plan across life-cycle (Dosing CDP)
 - Multiplicative considerations: molecule mode of action, indication specific efficacy need, patient tolerance/quality-of-life, CMC feasibility, cost and speed
 - Cohesive cross-functional partnership is essential
- Modeling-based approaches play an integral role in drug development and dose selection
 - Effectively integrate totality of evidence (PK, biomarker, efficacy, safety)
 - Leverage knowledge/data across molecules
 - Provide mechanistic insight and dosing projections
 - Guide study design and early go/no-go decision making potential
 - Wide range of modeling approaches are available (empirical, mechanistic, artificial intelligence).
 Selection of approach should be science/data driven and fit-for-purpose.



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Lunsumic

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pertuzumab for injection Amit Garg Bert I um



2022 Offsite (Monterey Bay, CA)

Ipatasertib

PERJETA

Angelica Quartino

Nageshwar Budha

Qi Liu Luna Musib **Bill Poland Russ Wada** Rui Zhu

Amita Joshi Chunze Li Samantha Roberts

Genentech Clin Pharm Team

Project Teams Internal Collaborators External Collaborators Investigators Patients

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Bei Wang

Justin Wilkins



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Using Pharmacokinetic-Pharmacodynamic Modeling and Simulation to Understand Dose and Exposure-Response Relationships for Adverse Reactions

Scott Van Wart, M.S., Ph.D Enhanced Pharmacodynamics, LLC, Buffalo, NY

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Scott Van Wart

I have the following relevant financial relationships to disclose:

I am the founder and Chief Scientific Officer of Enhanced Pharmacodynamics, LLC (ePD)

ePD consults with or has received research funding from various companies including: Adagio, Arcturus, AstraZeneca, Biogen, Boston Biomedical, Daiichi Sankyo, EMD Serono, Elevar Therapeutics, Eli Lilly, Enlaza Therapeutics, Enliven Therapeutics, EUSA, Genentech, Genmab, Glenmark, GSK, Halozyme, ICPD, Ionis Pharmaceuticals, Ironwood Pharmaceuticals, Jazz Therapeutics, Leyden Labs, Merck, Nektar Therapeutics, Nurix Therapeutics, Nutcracker Therapeutics, Oncoceutics, Praxis Therapeutics, PharmaEssentia, Roche, Sarepta Therapeutics, Seagen, Servier, Takeda, Teva, Tubulis, UCB Pharma, Veradermics, Via Nova, Windtree Therapeutics, and Zogenix.

Introduction

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- Exposure-response PK-PD analyses for safety are a critical component of model-based drug development
 - Used to provide decision-support criteria initially to guide dose escalation and later optimization during Phase 1 of development and to identify the RP2D
 - Safety AE and efficacy PK-PD models can be linked together to determine realistic patient dropout rates when performing simulations of ORR, OS and PFS
 - Can be used to understand potential impact of combination therapy or prophylactic use of other concomitant medications (e.g., GCSF) to counteract side effects
- This presentation will provide a few examples of how we have been asked to use PK-PD modeling and simulation for safety AE to help support dose regimen selection for clinical trials







Combining Logistic Regression Analysis of Safety AE Data with PK-PD Model for Tumor Growth to Optimize Dose

Review Totality of Safety AE Data





 Review totality of treatment-emergent AE (TEAE) classification data across clinical trials, and if possible by Body System / Organ Class or Preferred Term

| AE Classification | All Clinical Studies Combined (N=1000) | | | |
|--|---|--------------|-------|--|
| | No. Events | No. Patients | Rate | |
| All TEAE (any grade) | 1623 | 797 | 79.7% | |
| All TEAE of Grade 3+ | 324 | 297 | 29.7% | |
| TEAE of Blood and Lymphatic System Disorders (any grade) | 412 | 375 | 37.5% | |
| TEAE of Blood and Lymphatic System Disorders (Grade 3+) | 212 | 115 | 11.5% | |
| TEAE of Thrombocytopenia (any grade) | 512 | 285 | 28.5% | |
| TEAE of Grade 3+ Thrombocytopenia | 112 | 61 | 6.1% | |
| TEAE of Neutropenia (any grade) | 217 | 179 | 17.9% | |
| TEAE of Grade 3+ Neutropenia | 84 | 54 | 5.4% | |

Note: Rate is the percentage of the population with at least one TEAE of the given category

Exploratory PK-PD Analysis for Safety AE Data





- Boxplots can be used to show dose- and exposure-response relationships
- Examining different PK exposure metrics can help determine if there is signal and which metric might be most predictive prior to running the logistic regression



PK-PD Models for Efficacy and Safety AE

PK-PD Model for TGI

Logistic Regression PK-PD Model for Safety AE

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Predicted probability of Grade 3+ TEAE plotted vs. observed incidence rate within each AUC quantile

PK-PD Simulations to Optimize Dose







Note: Median and 90% PI shown for 500 virtual subjects. Dashed green line denotes 30% decrease in sum of diameters relative to baseline (solid black line).







Longitudinal PK-PD Models for Myelosuppression to Inform Dose Regimen and Clinical Trial Design

Comparison of Myelosuppression for Vyxeos[®] vs. Standard 7+3 Therapy





- Vyxeos[®] (CPX-351) is a liposomal encapsulation of daunorubicin (44 mg/m²) and cytarabine (100 mg/m²) approved for newly diagnosed therapy-related AML or AML with myelodysplasiarelated changes in adults and pediatric (aged ≥ 1 year) patients
- Population PK and PK-PD models were developed to characterize the impact of daunorubicin and cytarabine on ANC and platelets to compare liposomal Vyxeos[®] to standard 7+3 therapy

PK-PD Models for Myelosuppression for Vyxeos[®] vs. Standard 7+3 Therapy







Wang Q et al. S. Population PK-PD modeling of myelosuppression in patients with hematologic malignancies for CPX-351 and standard-of-care 7+3 therapy. Poster presented at 60th ASH meeting, 2018.

Cook S et al. Population PK-PD modeling of chemotherapy-induced neutropenia and thrombocytopenia in patients treated with CPX-351. Poster presented at ACoP9 meeting, 2018.



Friberg LE, et al. J Clin Oncol. 2002;20(24):4713-4721.

Neutrophil Model GCSF PK with target-mediated disposition



(10-10-1-0.01-0.001-0.0001-0.00001-0 576 1,152 1,728 2,304 Hours

100-

Melhem M, et al. Br J Clin Pharmacol. 2018;84(5):911-925.

Example Model Outputs for PD Effects



Application of PK-PD Models for Myelosuppression to an ADC







Time Since First Dose (week)

1.00

0.10

0.01





ADC dosed Q3W x 3



Neutrophil Model GCSF PK with target-mediated disposition



ADC dosed Q3W x 3



Simulations for Impact of Dose Interval and GCSF on ADC Dosing Regimen













Predicting the Impact of Combination Therapy on Myelosuppression to Select Starting Dose for Clinical Trials

Impact of Combination Vyxeos[®] and Venetoclax on Myelosuppression





- Venetoclax (B-cell lymphoma 2 inhibitor) is approved in combination with low-dose cytarabine for patients with AML who are ineligible for IC
- Semi-mechanistic PK-PD models for both ANC and platelet counts were developed to characterize the myelosuppressive effects of Vyxeos[®] and venetoclax monotherapy
 - Vyxeos[®] pop PK model was used along with mean PD data for first induction cycle from 3 clinical studies conducted in AML patients
 - Venetoclax pop PK model and mean PD data were obtained from literature
- Goal was to predict the safety profile and to recommend starting dose for low-intensity therapy of Vyxeos[®] plus venetoclax

Impact of Combination Vyxeos® and Venetoclax on Myelosuppression





PK-PD Model Fitting for CPX-351



Each unit of CPX-351 contains 1 mg cytarabine and 0.44 mg daunorubicin. ANC, absolute neutrophil count.



| | ANC | | Platelet | |
|--|-------------------|------------|---------------------|------------|
| Parameter | Final Estimate | RSE (%) | Final Estimate | RSE (%) |
| Circ ₀ (10 ⁹ /L) | 4.45 | 8.9 | 203 | 12.6 |
| MTT (h) | 66.6 | 5.0 | 87.2 | 9.0 |
| γ | 0.0278 | 8.1 | 0.258 | 18.5 |
| I _{max} | 1 | fixed | 0.313 | 11.2 |
| IC ₅₀ for CPX-351 (μM) | <mark>295</mark> | 8.96 | <mark>0.0109</mark> | 50.2 |
| IC ₅₀ for Venetoclax (µM) | <mark>120</mark> | 18.0 | <mark>16.5</mark> | 8.7 |

Liang Y, Cook SF et al. Population PK-PD Modeling of Neutrophil and Platelet Count for Lower-Intensity Therapy of CPX-351 Combined With Venetoclax in Acute Myeloid Leukemia. Poster presented at ASH Meeting, 2023.

Simulations to Select Starting Vyxeos[®] Dose for Phase 1b Study





20 units/m² Vyxeos[®] on Days 1 and 3; 400 mg Venetoclax QD until Day 21



- Simulations used to determine starting dose for LiT of Vyxeos[®]/venetoclax for Ph 1b study in newly diagnosed AML patients unfit for IC
- Targeted ≥ 50% of patients to recover above Grade 4 at end of 28-day cycle

Liang Y, Cook SF et al. Population PK-PD Modeling of Neutrophil and Platelet Count for LiT of CPX-351 Combined With Venetoclax in AML. Poster at ASH Meeting, 2023.







- These examples illustrate how PK-PD modeling and simulation can be used to support dose optimization and inform the design of clinical trials
- A wide range of PK-PD modeling approaches can be used to characterize safety AE data and the same data can in fact be modeled multiple ways (e.g., categorical vs. continuous timecourse models)
- Innovative approaches leveraging literature data or other published PK-PD models can provide tremendous value during drug development to help improve patient outcomes





MODERATOR

Stacy S. Shord, PharmD U.S. Food and Drug Administration

INTRODUCTORY SPEAKER

Jin Y. Jin, PhD *Genentech*

INTRODUCTORY SPEAKER

Scott Van Wart, PhD *Enhanced Pharmacodynamics*

ADDITIONAL PANELISTS

Youwei Bi, PhD U.S. Food and Drug Administration

Cara Rabik, MD, PhD U.S. Food and Drug Administration

W. Douglas Figg, PharmD *National Cancer Institute*

Julia Maues Patient Centered Dosing Initiative