#### **ADC:** Antibody-Drug-Conjugates



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## Clinical Pharmacology Challenges of ADCs in Early Drug Development

#### *Case study: Tusamitamab Ravtansine*

GPCO-Strasbourg October 2023

Laurent Nguyen

## **Tusamitamab Ravtansine (SAR408701)**





DAR (average) = 4

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## **Tusamitamab Ravtansine: Mechanism of Action**





DM4

S-Methyl-DM4 (Me-DM4) same potency

Lysine SPDB-DM4

Based on Erickson et al. Cancer Res. – April 15, 2006.

mAb-SPDB-DM4

## **METABOLIC PATHWAYS OF PAYLOAD (DM4)**





## **Clinical Pharmacology Challenges in Early Drug Development**



#### Bioanalytical consideration

- Because of ADC structure: Large and small molecules to be characterized in plasma
- Payloads: low circulating levels
- IMP is a mixture of different DAR species (DAR 0 to 8)

### What can be measured ?

Multiple analytes need different bioanalysis tools



## What can be measured ?

For Tusamitamab Ravtansine,

Proportion of individual DAR species in plasma were quantified by LC-MS/MS-high resolution

#### Individual DAR<sub>i</sub>





#### Proportion of individual DAR species

## **Clinical Pharmacology Challenges During Drug Development**



#### Bioanalytical consideration

- Because of ADC structure: Large and small molecules to be characterized in plasma
- Payloads: low circulating levels
- IMP is a mixture of different DAR species (DAR 0 to 8)

# PK characterization & modeling

- What are the PK characteristics of each component ?
- What are the PK variabilities ?
- How can we model all entities ?
- How to integrate mechanistic considerations ?

## **Integrated multiple analytes population PK model**

FIH study: 250 pts IV doses: 5 to 190 mg/m<sup>2</sup> Q2W, Q3W



#### ADC (DARi) model

- 2 compartments distribution
- Linear clearance

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- DAR1 to DAR8 species represented, same distribution (V<sub>c</sub> and V<sub>p</sub>) and clearance parameters (Q and CL<sub>ADC</sub>)
  - k<sub>dec,i</sub>: conversion of higher DAR to lower DAR successive deconjugations, modelled as an irreversible first-order process in central compartment
- **F**<sub>DAR,i</sub>: fraction of each DAR species in the administered solution

#### **NAB** (naked antibody)

 DARO species, same V<sub>c</sub>, Q and V<sub>p</sub> but specific CL<sub>NAB</sub> value  $\begin{array}{c} \mathsf{CL}_{\mathsf{ADC}} & \mathsf{Q} \\ \mathsf{ADC} & \mathsf{(V_c)} \end{array} \xrightarrow{\mathsf{Q}} \\ \mathsf{ADC} & \mathsf{(V_p)} \end{array}$ 

ADC is eliminated by two processes:

- Deconjugation releasing payload until NAB formation
- Global elimination reflecting systemic and/or cellular proteolysis

## **Integrated multiple analytes population PK model**



#### **DM4**

- 1 compartment distribution, linear elimination
- Each DAR≥1 deconjugation process was assumed to contribute to DM4 formation by releasing one DM4 molecule.

#### Me-DM4

- 1 compartment distribution, linear elimination
- Me-DM4 formed from DM4 elimination
- Fit simultaneoulsy of ADC, NAB, DM4,

Me-DM4 PK data, proportion of individual DAR species and average DAR

## Model prediction vs. observed data

2.00e4

1.80e4

1.60e4

1.40e4

1.20e4

Naked Ab

Cycle 1 data

Prediction corrected concentration (ng.Eq/mL)



DM4









## **Integrated multiple analytes population PK model**

<u>Final strue</u>	ctural model paran	neters:	
Fixed effects		- SD of the random	
Parameter	Estimate (RSE%)	effect, $\omega_p$ (RSE%)	
CL <sub>ADC</sub> (L/day)	0.392 (3)	46.9% (5)	
V <sub>c</sub> (L)	3.37 (2)	24.5% (5)	
<b>Q</b> (L/day)	0.543 (5)	52.9% (8)	
$V_p$ (L)	2.54 (5)	60.5% (8)	
<b>k<sub>dec8</sub></b> (/day)			
<b>k<sub>dec7</sub> (/</b> day)	0.938 (4)		
<b>k<sub>dec6</sub> (/</b> day)			
<b>k<sub>dec5</sub> (/</b> day)	0.751 (3)		
<b>k<sub>dec4</sub> (/</b> day)	0.525 (4)	20.2 (8)	
<b>k<sub>dec3</sub></b> (/day)	0.340 (4)		
<b>k<sub>dec2</sub></b> (/day)	0.181 (3)		
<b>k<sub>dec1</sub> (/</b> day)	0.0565 (2)		
CL <sub>NAB</sub> (L/day)	0.408 (3)	34.5 (6)	
CL <sub>DM4</sub> (L/day)	240 (3)	36.5 (6)	
<b>CL<sub>MeDM4</sub></b> (L/day)	0.256 (5)	65.4 (6) CL /F	-24 81/6
FR <sub>MeDM4</sub>	0.0107 (5)	72.3 (5)	M4-24.0 L/I



- ✓ Linear elimination (no TMDD)
- ✓ Stationnary clearance (no time-dependency)
- $\checkmark$  Low inter-occasion variability
  - (CV = 12% for ADC to 22% for MeDM4)

## **Individual DAR PK**

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DAR	Proteolytic clearance (L/d)	Deconjugation clearance (L/d)	Global clearance (L/d)
DAR≥6	0.392	3.16	3.55
DAR5	0.392	2.53	2.92
DAR4	0.392	1.77	2.16
DAR3	0.392	1.15	1.54
DAR2	0.392	0.611	1.00
DAR1	0.392	0.190	0.582
DAR0	0.408	-	0.408





#### batch-to-batch variability on ADC



## **Clinical Pharmacology Challenges During Drug Development**





#### Bioanalytical <u>consideration</u>

- Because of ADC structure: Large and small molecules to be characterized in plasma
- Payloads: low circulating levels
- IMP is a mixture of different DAR species (DAR 0 to 8)

#### PK characterization <u>& modeling</u>

- What are the PK characteristics of each component ?
- What are the PK variabilities ?
- How can we model all entities ?
- How to integrate mechanistic considerations ?

#### Intrinsic factors

- Sources of PK variabilities ?
- subpopulations at risks ?
- Is BSA dose normalized justified ?
- <u>Extrinsic factors:</u> <u>drug-drug interactions</u>
- Payload released can modulate enzyme and transporters mediated **DDIs**
- Perpetrator or victim

## Which intrinsic factors influence the PK ?

Covariate Population PK analysis from 248 pts (FIH study)



Five baseline covariates identified:

- > BSA
- Albuminemia
- Tumor size (tumor burden:TMBD)
- Circulating CEA (SHED)
- ➢ Gender

## Influence of covariates (5th and 95th percentiles) on typical population drug exposure for ADC, DM4 and MeDM4

(a)	SAR408701 steady-sta Base = 345	te AUCtau (μg.day/mL) μg.day/mL	(b)	DM4 steady-state AUCtau (ng.day/mL) Base = 5.6 ng.day/mL		(C) MeDM4 steady-state AUCtau (ng.day/mL) Base = 60 ng.day/mL			
	100 250	400 550	_	0 5	0 5 10			ò	100 200 300
AUCta	u 173 μg.day/mL	534 μg.day/mL	AUCtau	3.5 ng.day/mL		10 ng.day/mL	AUCtau	19 ng.day/mL	243 ng.day/mL
тмв	<b>D</b> 194 mm (-17.7%)	18.0 mm (+39.5%)	BSA	1.45 m² (-18%)		2.20 m² (+21.1%)	ALB	45.0 g/L (-37.2%)	30.0 g/L (+130.7%)
AL	<b>B</b> 30.0 g/L (-29.5%)	45.0 g/L (+19.7%)	SEX	Male (0%)		Female (+27.2%)	SHED	1866 pg/mL (-22.1%)	2254238 pg/mL (+20.7%)
BS	<b>A</b> 2.20 m <sup>2</sup> (-2.9%)	1.45 m² (+2.7%)	TMBD	194 mm (-7.9%)		18.0 mm (+14.2%)	BSA	1.45 m² (-18%)	2.20 m² (+21%)
SHE	D 2254238 pg/mL (-1.8%)	1866 pg/mL (+1%)	ALB	30.0 g/L (-12.9%)		45.0 g/L (+7%)	TMBD	18.0 mm (-23%)	194 mm (+14.1%)
SE	X Female (0%)	Male (0%)	SHED	2254238 pg/mL (-0.7%)		1866 pg/mL (+0.3%)	SEX	Female (0%)	Male (0%)

## Is BSA based dosing justified for Tusa ?

#### SAR408701 steady-state AUCtau

BSA (m<sup>2</sup>) quartiles [1.25, 1.63] [ [1.63, 1.80] [ [1.80, 1.97] [ [1.97, 2.66]





- ✓ BSA: significant covariate of pop PK model (ADC CL and Vc)
- ✓ BSA-based dosing avoids overexposure in low BSA group and under-exposure in high BSA group

## **Covariates effect: comparison with other ADCs**

DRUG	STRUCTURAL MODEL	COVARIATES ON ADC	COVARIATES ON PAYLOAD	DAR	CLEARANCE
<u>Gemtuzumab</u> ozogamicin	2 comp (T-ADC); 2 comp (payload) separated	CL: BW (fixed 0.75), DOSE, ALB Vc: BW (fixed 1), DOSE, ALB, SEX Vp: %Target_expression, blast_count, combination	None reported	Not included	Combined linear + time-dependent CL exp(-kdes) function
<u>Brentuximab</u> <u>vedotin</u>	3 comp (ADC) + 2 comp (payload) combined	CL, Q2, Q3: BW V1: BW, SEX V2, V3: BW	CL, Vc, Q, Vp: BW (fixed)	DAR = DAR.[a+(1- a).EXP(-b.TIME)]	Linear CL
<u>Trastuzumab</u> emtansine	2 comp (ADC) (payload < LOQ)	CL: BW, SHED, ALB, TMBD, Baseline_drug_concentration, ASAT Vc: BW	None reported	Semi-mechanistic model with interspecies first-order transfer rate constant	Linear (if 2 comp model) or non- linear CL (if semi- mechanistic model)
<u>Inotuzumab</u> ozogamicin	2 comp (ADC) (payload < LOQ)	CL: BSA, disease_ subtype, comedication, %blast Vc: BSA	None reported	Not included	Combined linear + time-dependent CL
<u>Polatuzumab</u> <u>vedotin-piiq</u>	2 comp (ADC) + 2 comp (payload) combined	CL: BW, SEX, ALB, combination, B_cell_count, TMBD, treatment_naive_status VC: BW, SEX, RACE, treatment_naive_status Q: BW Vp: BW	Formation_fraction: BW, SEX, treatment_naive_status, combination, hepatic_imp, ECOG, ALB	Not included	Combined non- specific linear time- dependent CL + linear time- dependent exponentially declining CL + non-linear CL
<u>Enfortumab</u> <u>vedotin</u>	3 comp (ADC); 2 comp (payload) separated	CL: BW, AGE, Hb, SEX, TMBD Q2, Q3: BW V1: BW, SEX, TMBD V2: BW V3: BW, tumor_type	CL: BW, ALB, ECOG, Hb, BILI Vc: BW, ALB Q: BW Vp: BW, ALB, Hb, RACE, GENDER	Not included	Linear CL
<u>Trastuzumab</u> <u>deruxtecan</u>	2 comp (ADC) + 1 comp (payload) combined	CL: BW, ALB, country, SEX, TMBD Vc: BW, SEX Vp: country	CL: comedication, AST, BILI, BW Vc: AGE, formulation	Not included	Linear CL
<u>Belantamab</u> mafodotin-blmf	2 comp (T-ADC) + 2 comp (ADC) + 1 comp (payload) combined	CL: BW, ALB, SHED, IgG, DOSE, study Vc: BW, ALB, SEX, study Vp: DOSE	Vc: SHED, IgG	DAR=DAR0*EXP(- RATE*TAD)	Combined linear + time-dependent CL sigmoid function
<u>Tisotumab</u> vedotin-tftv	2 comp (ADC) + 2 comp (payload) combined	CL: BW, ALB, SEX Vc: BW, ALB, SEX Q: BW Vp: BW	CL: BW, ALB, eGFR, Tumor_type, ECOG, hepatic_imp, TMBD Vc: BW, ECOG, ALB ktr: BW, AGE	DAR=1+3. EXP(- beta*TAD)	Combined linear + non-linear CL

- BSA or bodyweight are systematically relevant covariates
- Albumin, tumor burden and circulating target (sheding) are also relevant covariates for <u>both ADC & payloads</u>
- Covariates effects are of limited impact and do not require dose adjustment

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from Liu SN and Li C. CCP (2021) 87:743-765

## **Clinical Pharmacology Challenges During Drug Development**



#### Bioanalytical <u>consideration</u>

- Because of ADC structure: Large and small molecules to be characterized in plasma
- Payloads: low circulating levels
- IMP is a mixture of different DAR species (DAR 0 to 8)

- PK characterization & modeling
- What are the PK characteristics of each component ?
- What are the PK variabilities ?
- How can we model all entities ?
- How to integrate mechanistic considerations ?

#### Intrinsic factors

- Sources of PK variabilities ?
- subpopulations at risks ?
- Is BSA dose normalized justified ?

#### <u>Extrinsic factors:</u> drug-drug interactions

- Payload released can modulate enzyme and transporters mediated **DDIs**
- Perpetrator or victim

#### □ <u>PK/PD</u>

- Which entity best correlates with safety and efficacy endpoints?
- What are the relevant PK metrics ?
- How to model PK/PD relationships ?

### What is the driver(s) of efficacy (tumor size decrease) ?



#### Nsq NSCLC expansion cohorts from FIH

**NSQ-NSCLC** moderate expressors

**NSQ-NSCLC** high expressors

#### ADC concentrations is the best driver of tumor Size dynamics



✓ No effect of DM4 or Me-DM4 PK

• 
$$\frac{d(TS)}{dt} = [k_{ge}, TS(t)] \cdot [1 - k_{kill} \cdot C_{SAR408701}(t) \cdot e^{-lambda(t)}]$$





## What is the driver(s) of toxicity ?

Exposure versus safety multivariate analyses (211 pts - 5 to 190 mg/m<sup>2</sup> Q2W – FIH study)

Corneal event main dose-limiting toxicity (26% of grade  $\geq$  2)



> Corneal event is mainly driven by cycle 1 ADC exposure

No or limited Contribution of payload

## **PK/PD: comparison with other ADCs**

Cancer Chemotherapy and Pharmacology (2021) 87:743-765 https://doi.org/10.1007/s00280-021-04250-0

**REVIEW ARTICLE** 

Clinical pharmacology strategies in supporting drug development and approval of antibody–drug conjugates in oncology

Stephanie N. Liu<sup>1</sup> · Chunze Li<sup>1</sup>

.."for most of the seven approved ADCs, the efficacy endpoints appear to correlate best with ADC conjugate compared to that of unconjugated payload.

For **safety** outcomes, while ADC exposures were often correlated with AEs, **unconjugated payload exposures may also be important for certain AEs**"

#### Exposure-safety and exposure-efficacy analyses for tisotumab vedotin for patients with locally advanced or metastatic solid tumors

Chaitali Passey<sup>1</sup> | Jenna Voellinger<sup>2</sup> | Leonid Gibiansky<sup>3</sup> ◎ | Rudy Gunawan<sup>2</sup> | Leonardo Nicacio<sup>2</sup> | Ibrahima Soumaoro<sup>1</sup> | William D. Hanley<sup>2</sup> | Helen Winter<sup>4</sup> | Manish Gupta<sup>1</sup>

CPT PsP 2023:1-12

ORR

#### Occular Grade 2+



## Conclusions







#### Bioanalytical <u>consideration</u>

- Need several bioanalytical tools to measure different components
- Low circulating payloads need sensitive assays

#### PK characterization & modeling

- Population PK analysis enables multiple analytes integration with mechanistic considerations
- Useful to characterize the PK of all entities and to draw CMC specifications (e.g. DAR variability)

#### Intrinsic factors

- Payload PK is much more variable than ADC PK
- Standard covariates (ALB, Tumor burden, circulating target) are commonly identified but of limited impact on ADC and paylaod exposure
- BSA or body weight are usually relevant covariates
- Impact of immunogenicity ?

#### drug-drug interactions

• IVIVE or PBPK are useful to predict DDI mediated by payloads

#### D PK/PD

- Difficult to handle multiple analytes effects and to determine the best driver
- Combination of different approaches may help (E-R, longitudinal PK/PD modeling, PBPK, QSP)
- QSP modeling is likely meaningful (Scheuher B. et al. JPP, 2023)
- Impact of immunogenicity ?

#### **ORIGINAL PAPER**

#### Towards a platform quantitative systems pharmacology (QSP) model for preclinical to clinical translation of antibody drug conjugates (ADCs)

Bruna Scheuher<sup>1,2</sup> · Khem Raj Ghusinga<sup>1</sup> · Kimiko McGirr<sup>1</sup> · Maksymilian Nowak<sup>1</sup> · Sheetal Panday<sup>1</sup> · Joshua Apgar<sup>1</sup> · Kalyanasundaram Subramanian<sup>1,3</sup> · Alison Betts<sup>1,2</sup>



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Journal of Pharmacokinetics and Pharmacodynamics (2022) 49:381-394 Pouzin C. et al. JPP 2022; 49:381:394 https://doi.org/10.1007/s10928-021-09799-0 **ORIGINAL PAPER** Integrated multiple analytes and semi-mechanistic population pharmacokinetic model of tusamitamab ravtansine, a DM4 anti-CEACAM5 antibody-drug conjugate Clemence Pouzin<sup>1,2</sup> · Leonid Gibiansky<sup>3</sup> · Nathalie Fagniez<sup>1</sup> · Mustapha Chadjaa<sup>4</sup> · Michel Tod<sup>2</sup> · Laurent Nguyen<sup>1</sup> Received: 1 September 2021 / Accepted: 20 December 2021 / Published online: 15 February 2022 C The Author(s) 2022 Revised: 12 January 2022 Accepted: 24 January 2022 Received: 9 November 2021 DOI: 10.1002/psp4.12769 Pouzin C. et al. CPTpsp 2022;11(3):384-394. ARTICLE Covariate analysis of tusamitamab ravtansine, a DM4 anti-CEACAM5 antibody-drug conjugate, based on first-in-human study

Clemence Pouzin $^{1,2} \mid Michel Tod^2 \mid Mustapha Chadjaa^3 \mid Nathalie Fagniez^1 Laurent Nguyen^1$ 

Prediction of CYP down regulation after tusamitamab ravtansine administration (a DM4-conjugate), based on an in-vitro-in-vivo extrapolation approach CPT 2023 accepted

Clemence Pouzin<sup>1</sup>, Donato Teutonico<sup>1</sup>, Nathalie Fagniez<sup>1</sup>, Samira Ziti-Ljajic<sup>1</sup>, Anne Perreard-Dumaine<sup>2</sup>, Magalie Pardon<sup>3</sup>, Sylvie Klieber<sup>4</sup>, Laurent Nguyen<sup>1</sup>



# • Thank you •



# BACK UP



### **Clinical Pharmacology considerations for ADC development**



Inform platform PK characteristics and dose strategy

From Liu SN, CCP (2021)87:743-765

## **IVIVE-PK model to predict CYP down regulation**

#### In vitro data show ٠

- **Down regulation** of CYP3A4, 1A2 • and 2B6 by Me-DM4 and DM4 observed in human hepatocytes
- Mechanism based inhibition (MBI) • of CYP3A4 by DM4 in human liver microsomes
- Simulation of CYP abundance time course reduction
- **Prediction of MDZ AUC ratio** (*PBPK simulation*)  $\geq$



### **IVIVE PK simulations to predict CYP3A4 abundance decrease**



- Transient effect
- Less than 20% decrease in CYP3A4 abundance when considering extreme values (5<sup>th</sup> percentile)
- Expected AUC ratio for Midazolam CYP3A4 probe =1.14 (< 1.25) => no clinically relevant DDI effect on CYP3A4 substrate

### **IVIVE-PK model simulations to predict CYP down regulation**



- Below 10% decrease of CYP abundance predicted by IVIVE modeling
- > Expected AUC ratio for Midazolam CYP3A4 probe =1.14 (< 1.25) => no clinically relevant (DDI guidelines)

## Cycle 1 TED



### **ADCs: Clinical use**

	Name   Trade Name, Company	Antibody Target	Payload	Payload Target	DAR	Approved Indication
	Gemtuzumab ozogamicin   Mylotarg ® Pfizer	CD33	Calicheamicin	DNA	2-3	Acute Myeloid Leukemia
AL	Brentuximab vedotin   Adcetris ® Seagen	CD30	MMAE	Microtubule	4	HL and ALCL, PTCL, cHL
9	Inotuzumab ozogamicin   Besponsa ® Pfizer	CD22	Calicheamicin	DNA	5-7	Acute Lymphocytic Leukemia
	Moxetumomab pasudotox   Lumoxiti ® AstraZeneca	CD22	PE38	-	NA	Hairy Leukemia
MA	Polatuzumab vedotin   Polivy ® Roche	CD79b	MMAE	Microtubule	3.5	Diffuse Lymphoma
Ë	Belantamab mafodotin   Blenrep ® GSK	BCMA	MMAF	Microtubule	4	Multiple myeloma
	Loncastuximab tesirine   Zynlonta ® ADC	CD19	PBD dimer	DNA	2.3	Diffuse Lymphoma

	Trastuzumab emtansine <sup>1</sup>   Kadcyla ® Roche	HER2	DM1	Microtubule	3.5	Breast
	Enfortumab vedotin   Padcev ® Seagen	Nectin4	MMAE	Microtubule	3.8	Urothelial
ORS	Trastuzumab deruxtecan   Enhertu ® Daiichi	HER2	Dcd	DNA	7-8	Breast, Gastric, Lung <sup>2</sup>
M	Sacituzumab govitecan   Trodelvy ® Immunomedics	TROP2	SN38	DNA	7.6	Triple Negative Breast Cancer
	Cetuximab saratolacan <sup>3</sup>   Akalux ® Rakuten	EGFR	IRdye700DX	Cell membrane	1.3-3.8	Head & Neck
SOL	Tisotumab vedotin   Tivdak ® Seagen	TF	MMAE	Microtubule	4	Cervical
	Disitamab vedotin   Aldixi ® Remegen	HER2	MMAE	Microtubule	4	Gastric/ Gastroesophageal
	Mirvetuximab soravtansine   Elahere ® ImmunoGen	FRα	DM4	Microtubule	3.5	Ovarian, Fallopian, Peritoneal

Adapted from Zhiwen Fu et al. Nature 2023

1<sup>st</sup> ADC approved in 2000 for AML (Mylotarg®)

> 15 Approved ADCs

 Pipeline is exponentially growing with more than 100 clinical studies

ALCL= Anaplastic Large Cell Lymphoma; ALL = Acute Lymphocytic Leukemia; cHL= Classic Hodgkin Lymphoma; FR= Folate Receptor; HL = Hairy leukemia; PTCL = Peripheral T Cell Lymphoma; TF = Tissue Factor

### Key components of an ADC

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ADC evolution (from Beck et al. 2017)

2 3

Number of Drugs/Antibody

34

1 2 3 4 5 6 7 8

Number of Drugs/Antibody

## **Covariates**

#### Baseline continuous covariates

	Min	Max	Median
Age (years)	31	91	62
Body weight (kg)	36.0	138.0	69.0
Body surface area (m²)	1.25	2.66	1.80
Body mass index (kg/m²)	14.6	40.8	24.1
Creatinine clearance (mL/min)	35.7	213.3	90.3
Albumin (g/L)	24.0	48.0	39.0
Bilirubin (µM)	1.70	51.3	7.83
Total Protein (g/L)	53.0	89.7	72.0
ASAT (IU/L)	10.0	208	25.0
ALAT (IU/L)	5.00	166.0	18.5
Tumoral CEACAM5 expression (%)	0	100	70.0
Tumor Burden (mm)	11.0	339	84.0
SHED: Circulating CEA (pg/mL)	500	41227000	56530
HSCORE	0	300	210

#### **Baseline categorical covariates**

	Subclass
SEX	Male: N=156 Female: N=98
ETHNIC	Non Hispanic: N=224 Hispanic: N=30
RACE	Caucasians: N=209 Blacks: N=0 Oriental: N=45 Other: N=0
ECOG (Eastern Cooperative Oncology Group status)	PS 0: N= 87 PS 1: N=165 PS 2: N=1 PS 3: N=1
TUMOR TYPE	Breast: N=1 Colon/Rectum: N=93 Esophagus: N=1 Gastroesophageal: N=10 Lung: N=119 Pancreas: N=5 Stomach: N=25

## **Impact of renal impairement on Tusa PK ?**



#### Renal status based on eGFR (MDRD)

No difference in ADC and payloads exposure between Mild/moderate RI vs normal renal function patients

## **Impact of liver impairement on Tusa PK ?**



Hepatic status was defined according to the National Cancer Institute

Confounding effect of unbalanced covariates distribution between mild LI patients (higher tumor burden, lower albumin and higher circulating CEA) and normal patients. LI could be associated with disease severity or worsening.

#### Tumor Size dynamics was accurately characterized by ADC concentrations





- ✓ No added value of DM4 or Me-DM4 PK concentrations.
  - $\frac{d(TS)}{dt} = [k_{ge}.TS(t)] \cdot [1 k_{kill} \cdot C_{SAR408701}(t) \cdot e^{-lambda(t)}]$
- ✓ Trend for better response for higher CEACAM5 expressors.





## **1.** Tusamitamab ravtansine pop PK

#### > MODEL EVALUATION: OBSERVED DATA vs iPRED



#### Impact of covariates (ALB and Tumor burden) on ADC and MeDM4 PK



Observed data from Nsq NSCL cancer pts (n=62 pts)

- ✓ Inverse correlation btw ADC and MeDM4 exposure
- ✓ High ALB and low tumor burden reflect lower ADC proteolysis and/or deconjugation
  ♥ Higher ADC exposure and lower MeDM4 exposure
- ✓ high tumor burden and low ALB reflect high proteolysis and/or deconjugation
  ✤ Low ADC exposure and high MEDM4 exposure