Modified Toxicity Probability Intervals (mTPI),
Bayesian Logistic Regression Modeling (BLRM),
Continual Reassessment Method (CRM)
via EAST6.3.1

VS.

T-statistic (Tstat) Design via COMPASS for finding MTD

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JSM2016 01Aug2016

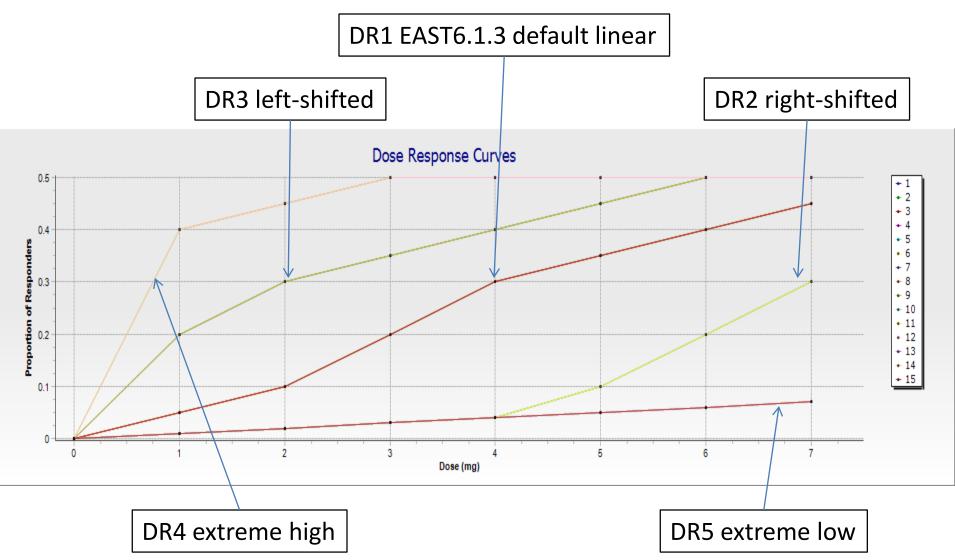


#### Motivation

- mTPI is appealing and popular for adaptive dose-finding of Maximum Toxicity Dose (MTD)
  - Easy to implement (fixed pre-stated algorithm; DR-model-independent)
  - Efficient as competitor DR-model-based designs (CRM, BLRM)
  - Better than traditional 3+3 design
- T-statistic design is appealing and popular with some for adaptive dose-finding of Target Dose
  - Easy to implement (requires simple calculation after each cohort; based on isotonic DR-model)
  - Efficient as competitor designs (Bayesian 4PL, Emax, NDLM)
  - Better than fixed-randomized designs



# Dose-Response Curves Simulated (MTD = dose with probability of response = 0.3)



### **Design Parameters**

- 7 doses (1,2,3,4,5,6,7)
- Total N=30 subjects
  - 10 sequential cohorts of 3 subjects each
- 1st cohort at Dose1
- Each subsequent cohort assigned a single dose per adaptive design
- Target Toxicity Level 0.3
- 10K simulations of each DR curve scenario



#### mTPI method

- Bayesian Posterior Probability that TRUE DLT rate lies in each of 3
  Toxicity Intervals
  - Under dosing: <0.25</p>
  - On-Target dosing: 0.25-0.35
  - Over dosing: >0.35
- Prior on TRUE toxicity probability at each dose ~ Beta(1,1)
- Applies up/down/stay rules for next dose based on posterior probabilities of being in each toxicity interval at current dose
- Over-Dosing Exclusion Rule
  - Prob(Pi>Pt|data) > 0.9999 [to yield full sample size]
    - Pi = Prob(toxicity at dose i)
    - Pt = Target probability of toxicity = 0.3
    - Similar results for EAST default Prob > 0.95 (not shown)
    - Prob>0.6 also assessed
- No Under-dosing exclusion rule used



### mTPI "optimization"

- 2 levels of early stopping
  - posterior probability required (0.95 and 0.80).
- 2 toxicity probability ranges:

- beta(1,2) prior to yield estimate of ~0.3 for probability of toxicity at each dose since the target toxicity level is 0.3
  - beta(1,1) also run; it yields estimate 0.5



### T-stat Parameters (1)

- T = (Pi-0.3)/sqrt((Pi\*(1-Pi)/n))
  - Pi = isotonic estimated proportion of toxicity at Dose i
- Dose escalation / de-escalation rules:
  - T < -2  $\rightarrow$  up 2 dose increments
  - $-2 \le T < -0.1$   $\rightarrow$  up 1 dose increment
  - $-0.1 \le T < 0.1 \implies$  repeat dose
  - $0.1 \le T < 2$   $\rightarrow$  down 1 dose increment
    - 2 ≤ T → down 2 dose increments



### T-stat Parameters (2)

- T = (Pi-0.3)/sqrt((Pi\*(1-Pi)/n))
  - Pi = isotonic estimated proportion of toxicity at Dose i
- Dose escalation / de-escalation rules:

```
T < -2 \rightarrow up 2 dose increments
```

$$-2 \le T < -1 \implies$$
 up 1 dose increment

$$-1 \le T < 1 \implies$$
 repeat dose

$$1 \le T < 2 \implies$$
 down 1 dose increment

- Early Stopping Via Post.Prob.(toxicity rate > 0.35)
  - Three cutoffs (0.5, 0.65, and 0.8)

NOTE: T-stat may have unfair advantage over dose esc. designs since it can skip a dose in extreme cases (i.e., |T|>2)

# BLRM Parameters (Bayesian Logistic Regression Modeling)

Toxicity Intervals

Under dosing: <0.25

Target toxicity: 0.25-0.35

Excessive toxicity: 0.35-0.45

Unacceptable toxicity: >0.45

- Prior Distribution: Bivariate Lognormal for the 2 logistic parameters
- Prior on Lowest Dose and MTD
  - Prob.(DLT) at D1 = 0.05
  - Estimated MTD = 4
  - # Beta Samples = 1000 (Direct Sampling; default settings)
- Probability(Overdosing) < 0.25</li>
- No Early Stopping



# BLRM "Optimization"

- 3 priors for logistic regression model, per Neuenschwander(2008)
- Doses with EWOC OD post.prob > 0.25,0.5, 0.8 evaluated



# CRM Parameters (Continual Reassessment Method)

- Target Probability of Toxicity = 0.3
- Toxicity Probability Upper Limit = 0.3
- Model Type = 1-parameter power and logistic
   Gamma(1,1) prior
- Default prior

```
Doses: D1 D2 D3 D4 D5 D6 D7 Prob(tox): 0.05 0.1 0.2 0.3 0.35 .4 0.45
```

 No EWOC, but early stopping if Prob(lowest dose toxicity rate > 0.3)>0.9

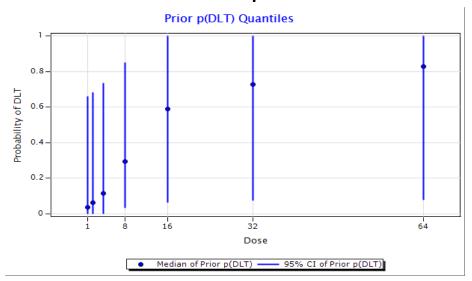


## CRM "Optimization"

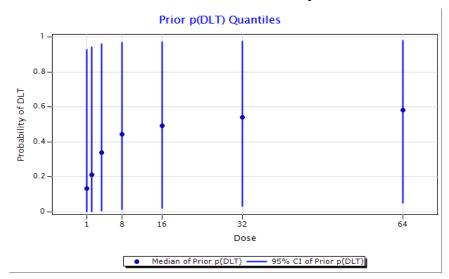
- Permit skipping doses since Tstat dose so
- Permit dose escalation if a prior subject experienced a toxicity
- Three priors (CRM1,2,3) chosen for similarity to the priors for BLRM plus a very-close-to-flat prior (CRM4) for one-parameter power model:
  - all use gamma(1,1) as prior for the power parameter
- 2 upper toxicity probability limits for the EAST default one-parameter logistic model



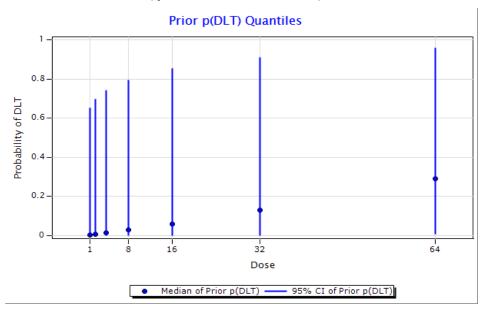
#### BLRM default prior



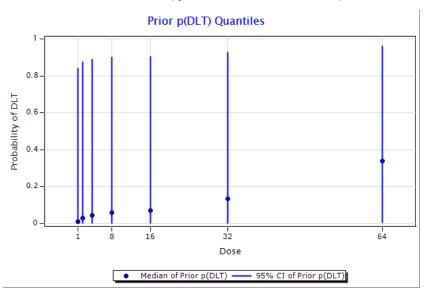
#### CRM default prior



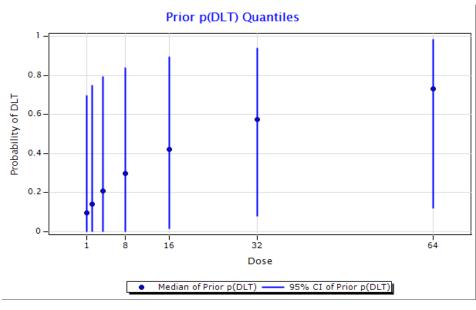
#### BLRM(prior1 "low")



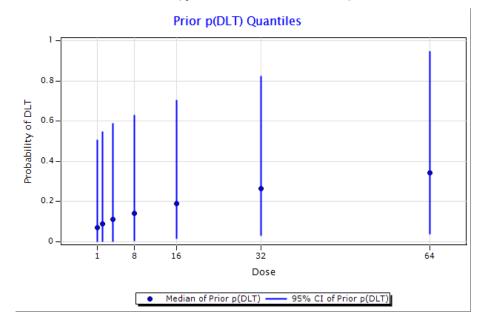
#### CRM(prior1 "low")



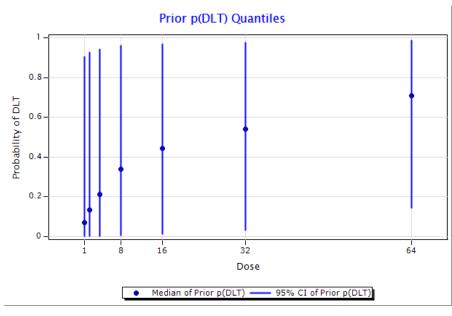
#### BLRM(prior2 "high")



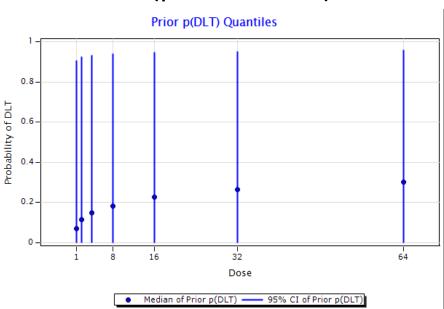
#### BLRM(prior3 "mid")



#### CRM(prior2 "high")



#### CRM(prior3 "mid")



# Remarks on mTPI vs BLMR vs CRM vs Tstat from simulations

- Tstat design looks like a competitor to mTPI, BLRM, CRM for toxicity dose-finding trials
  - Based on 4 Performance Criteria:
    - Probability of identifying correct target ID,
    - Probability of estimating MTD at or adjacent to correct MTD
    - probability of assigning subjects to doses > target (OD's),
    - # dose-limiting toxicities (DLT's) observed
  - Each of the 4 designs could be optimized better than the others <u>for</u>
     <u>particular individual</u> DR curves and/or <u>particular performance</u>
     <u>criteria</u>
- Indications are that Tstat is competitive with mTPI, BLRM, CRM in consideration of the spectrum of TRUE underlying DR curves simulated when the 4 performance criteria are combined with equal weights



# One Way to Rank the Designs Across the Five DR Curves & 3 Performance Criteria

- Weight each of 4 performance criteria 1:1:1:1 since 2 assess MTD estimation and 2 assess safety
- Compute "relative difference from optimal over all design scenarios" for each DR curve for each design:
  - (Max.%correct %correct)/(Max. Min.%correct)
  - (Max.%at\_near %at\_near)/(Max. Min.%at\_near)
  - 100 (Prob.Assgn>Tgt Min.Prob.)/(Max. Min.Prob.Assgn>Tgt)
  - 100 (Avg#tox Min.Avg#tox)/(Max. Min.Avg#tox)
- Then compute average across all DR curves & multiply by 100
- Values closer to 100 indicate closer to optimal design
- Values closer to 0 indicate closer to worst design
- Values = 50 indicate mid-way between optimal and worst designs



# Summary Scores Limited to Best 2 Designs of each type

- Wide range of values for each performance characteristic could unduly inflate / deflate summary scores
- Hence, choose best 2 designs of each type and re-compute summary scores based on only those 8 designs
- Each of the designs out-performs the other 3 designs for at least one performance criteria across all 5 DR curves or for at least one DR curve across all 4 performance criteria (next slide)
- Tstat Design performed best overall, but not by much, across ALL DR curves for average of all 4 performance criteria scores (next slide)

#### **Summary Scores Limited to Best 2 Designs of each type**

	0'	ver all D	R curv	es	over all 4 perf.char.							
		%at/										
design	%at	nexto	#OD	#DLTs	DR1	DR2	DR3	DR4	DR5	DR1-5		
Tstat2s3550_f	62	50	85	65	45	69	54	<u>93</u>	50	[65]		
Tstat1s3550_f	65	52	77	60	44	<u>72</u>	54	86	50	64		
mTPIr2s9p11_m	40	40	<u>100</u>	60	<u>100</u>	50	33	25	67	60		
BLRMpr2-s25_f	51	<u>65</u>	41	53	73	55	61	50	24	53		
BLRMdefault_f	20	40	53	<u>92</u>	90	53	33	37	33	51		
CRM_defltPW_m	60	60	0	40	0	50	<u>67</u>	75	33	40		
mTPIr5s7p11_f	<u>71</u>	47	17	20	19	18	59	36	<u>89</u>	38		
CRM_defltPW_f	61	57	14	14	23	28	17	63	67	37		

- Each of the design type out-performed the other 3 design types for at least one performance criteria across all 5 DR curves or for at least one DR curve across all 4 performance criteria (bold underlined results above)
- Tstat Design performed best, but not by much, across ALL DR curves for average of all 4 performance criteria scores [above]

#### Remarks on mTPI vs BLMR vs CRM vs Tstat

- Tstat design looks like a competitor to mTPI, BLRM, CRM for toxicity dose-finding trials
  - Based on Performance criteria:
    - Probability of identifying correct target ID,
    - Probability of estimating MTD at or adjacent to correct MTD
    - probability of assigning subjects to doses > target (OD's),
    - # dose-limiting toxicities (DLT's) observed
  - Each of the 4 designs could be optimized better than the others <u>for</u>
     <u>particular individual</u> DR curves and/or <u>particular performance</u>
     <u>criteria</u>
- Indications are that Tstat is competitive with mTPI, BLRM, CRM in consideration of the spectrum of TRUE underlying DR curves simulated when the 4 performance criteria are combined with equal weights



### Next Steps re: Tstat for toxicity dose-finding

- Consider Tstat in addition to traditional adaptive escalation designs as in EAST
- Consider other ranking mechanisms to compare the design performance characteristics
- Evaluate additional design configurations to optimize, e.g., enhancements in EAST6.4
- Consider Ivanova(2012) Bayesian Isotonic Adaptive Dose-Finding design vs mTPI, CRM, BLRM, Tstat

Other ?? [DISCUSSION: bolognese@cytel.com]



#### **REFERENCES**

- Ivanova A, Bolognese J, Perevozskaya I. Adaptive design based on T-statistic for dose-response trials. Statistics in Medicine, 2008 May 10;27(10):1581-92
- EAST6.3.1 User Manual, Cytel Inc., Cambridge, MA 2014
- COMPASS User Manual, Cytel Inc., Cambridge, MA, 2012
- Ivanova A, Xiao C, Tymofyeyev Y. Two-stage designs for Phase 2 dose-finding trials. *Statist. Med.* 2012; **31:**2872–2881
- Ji Y, Liu P, Li Y, and Bekele N (2010). A modified toxicity probability interval method for dose finding trials. Clinical Trials, 7:653-656.
- Neuenschwander B, Branson M, and Gsponer T (2008). Clinical aspects of the Bayesian approach to phase I cancer trials. Statistics in Medicine, 27:2420-2439.
- O'Quigley J, Pepe M, and Fisher L (1990). Continual reassessment method: A practical design for phase I clinical trials in cancer. Biometrics, 46:33-48.
- Bolognese JA, Patel N, Tymofyeyev Y, Perevozskaya I, Palmer J. T-Statistic-based Up&Down Design for Dose-Finding Competes Favorably with Bayesian 4parameter Logistic Design. Joint Statistics Meetings, Washington, DC, August 5, 2009. (invited presentation)



# SUPPLEMENTAL INFO FOLLOWS THIS SLIDE



#### mTPI cases simulated

mTPIdefault_f	mTPlr2s9p11_m
mTPIdefault_m	mTPIr2s9p37_f
mTPIdeflt37_f	mTPIr2s9p37_m
mTPIdeflt37_m	mTPIr5s7p11_f
mTPIr2s7p11_f	mTPIr5s7p11_m
mTPlr2s7p11_m	mTPIr5s7p37_f
mTPIr2s7p37_f	mTPIr5s7p37_m
mTPIr2s7p37_m	mTPIr5s9p37_f
mTPIr2s9p11_f	mTPIr5s9p37_m

("m" indicates MTD estimate via the method description in the reference; "f" indicates via isotonic regression fit)



## Tstat cases simulated

Tstat1f	Tstat2f
Tstat1m	Tstat2m
Tstat1s3550_f	Tstat2s3550_f
Tstat1s3550_m	Tstat2s3550_m
Tstat1s3565_f	Tstat2s3565_f
Tstat1s3565_m	Tstat2s3565_m
Tstat1s3580_f	Tstat2s3580_f
Tstat1s3580_m	Tstat2s3580_m
Tstat1s5080_f	Tstat2s5080_f
Tstat1s5080_m	Tstat2s5080_m

## BLRM cases simulated

BLRMdefault_f	BLRMpr2-s50_f
BLRMdefault_m	BLRMpr2-s50_m
BLRMpr1-s25_f	BLRMpr2-s80_f
BLRMpr1-s25_m	BLRMpr2-s80_m
BLRMpr1-s50_f	BLRMpr3-s25_f
BLRMpr1-s50_m	BLRMpr3-s25_m
BLRMpr1-s80_f	BLRMpr3-s50_f
BLRMpr1-s80_m	BLRMpr3-s50_m
BLRMpr2-s25_f	BLRMpr3-s80_f
BLRMpr2-s25_m	BLRMpr3-s80_m

## **CRM Cases simulated**

CRM_defltLG_f	CRM_p2modLG_f
CRM_defltLG_m	CRM_p2modLG_m
CRM_defltPW_f	CRM_p2modPW_f
CRM_defltPW_m	CRM_p2modPW_m
CRM_p1modLG_f	CRM_p3modLG_f
CRM_p1modLG_m	CRM_p3modLG_m
CRM_p1modPW_f	CRM_p3modPW_f
CRM_p1modPW_m	CRM_p3modPW_m

### BLRM Priors (Neuenschwander, 2008)

- bivariate normal prior for means of log(alpha) and log(beta) in the Bayesian logistic linear regression model:
  - mean(alpha) = logit(p-star)=log(0.3/0.7)=-0.847, where p-star is the target toxicity level

```
mean(beta) = 0, SD(alpha) = 2, SD(beta) = 1, correlation = 0
```

- setting prior probabilities of
  - (1) exceeding the minimum unacceptable toxicity proportion at the lowest dose, and
  - (2) falling below the maximum under-dosing toxicity proportion at the highest dose at, e.g., 0.05.
  - then deriving corresponding multivariate normal parameters.
  - For Prob(p1>0.6)=0.05 and Prob(pK<0.2=0.05), the corresponding 5 multivariate normal parameters (m1,m2,s1,s2,rho) are (-0.376, -0.466, 0.853, 0.931, -0.119).</li>
- Flatter prior: mean(alpha)=-1.025, mean(beta)=-1.091, SD(alpha)=0.893, SD(beta)=1.147, corr=-0.084.



### T-stat Dose-Stopping Rules

- Posterior probability (with beta(1,1) prior) that estimated toxicity at a dose > unacceptable toxicity level exceeds cutoff, then that dose and all higher doses no longer assigned
- Three cutoffs (0.5, 0.65, and 0.8) simulated for unacceptable toxicity level 0.35
- As a liberal criteria, cutoff 0.8 simulated for unacceptable toxicity level 0.5.
- Simulated for each of Tstat(1) and Tstat(2)
- Also considered NO dose-stopping
- MANY THANKS to Jaydeep Bhattacharyya for programming the early stopping into CytelSim



## CRM "Optimization"

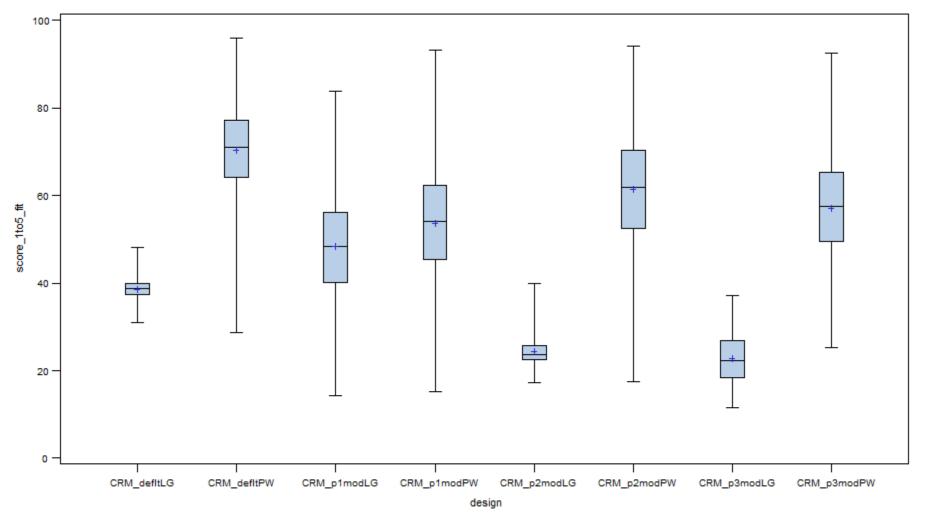
- Permit skipping doses since Tstat dose so
- Permit dose escalation if a prior subject experienced a toxicity
- Three priors (CRM1,2,3) chosen for similarity to the priors for BLRM plus a very-close-to-flat prior (CRM4) for one-parameter power model:
  - all use gamma(1,1) as prior for the power parameter

```
— D1 D2 D3 D4 D5 D6 D7
```

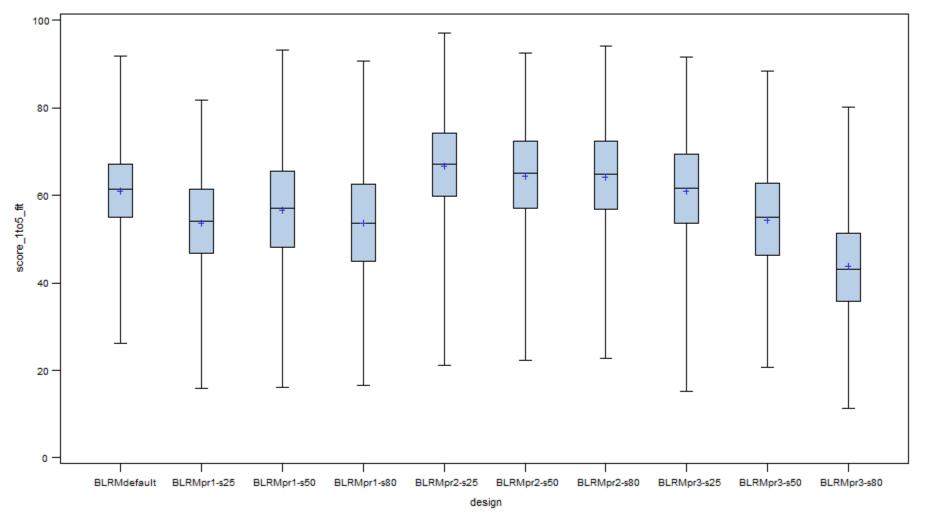
- Default prior 0.05, 0.1, 0.2, 0.3, 0.35, 0.4, 0.45
- Prior1 0.001, 0.005, 0.01, 0.015, 0.02, 0.05, 0.2
- Prior2 0.02, 0.05, 0.1, 0.2, 0.3, 0.4, 0.6
- Prior3 0.02, 0.04, 0.06, 0.08, 0.11, 0.14, 0.17
- 2 upper toxicity probability limits for the EAST default oneparameter logistic model
- 6 CRM configurations in all



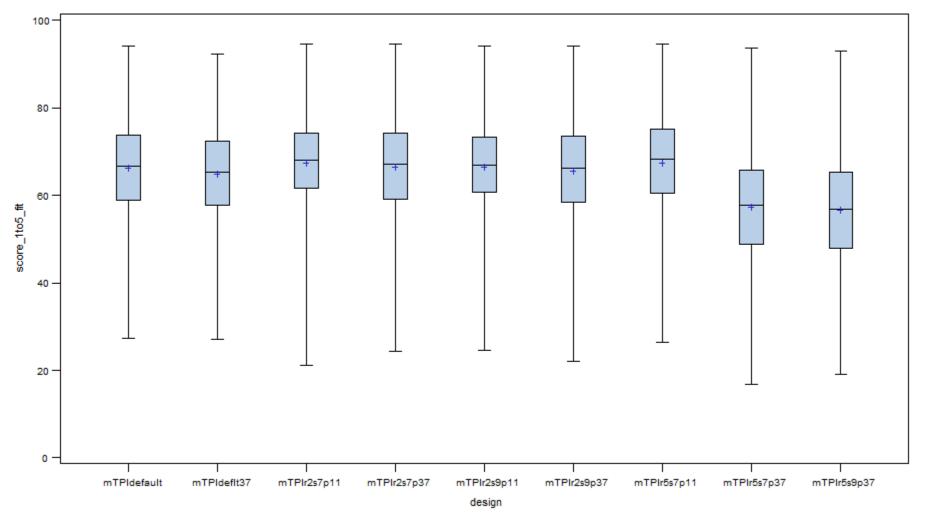
design=CRM

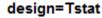


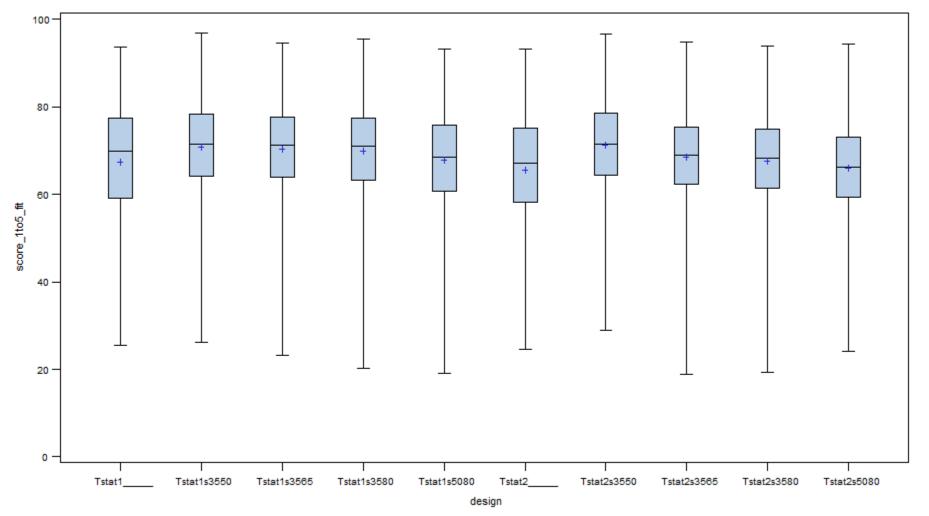
design=BLRM



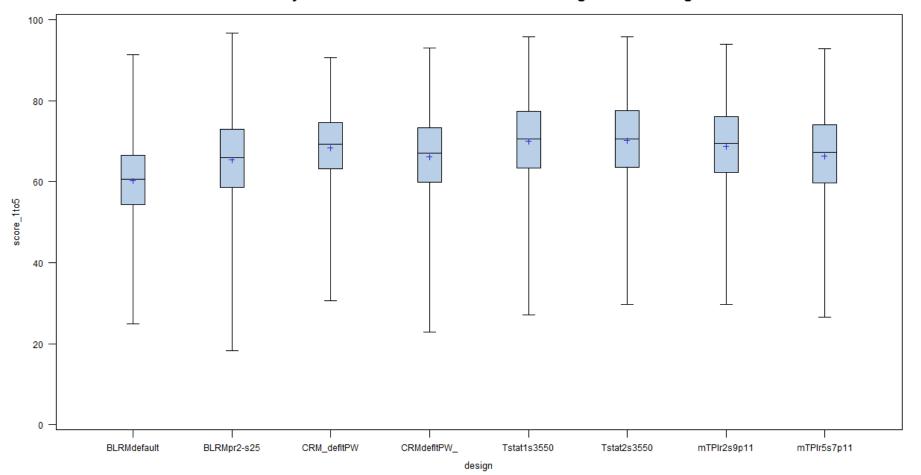
design=mTPI







summary score over all 5 DR curves - best 2 configurations / design



				T	st	at	a	s e	ea	Sy	' a	S	m	TF	)	to	ir	ηį	ole	en	ne	n.	t					
Number	T-Sta	atistic	Des	ign fo	r Tar	get To	oxicit	y Lev	el =	0.3				rr	nodif	y yell	ow-h	ighlig	ghted	cells	to m	odify	y desi	gn sp	ec's			
of											N	lumb	er of	Subje	ects (	Obse	rved											
Toxicities	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30
0	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
1	0	0	0	0	1	1	1	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
2	-1	0	0	0	0	0	0	0	1	1	1	1	1	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
3	-2	-2	-1	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	2	2	2	2	2	2	2	2	2	2
4		-2	-2	-1	-1	-1	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	2	2	2
5			-2	-2	-2	-1	-1	-1	-1	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	1
6				-2	-2	-2	-2	-1	-1	-1	-1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1
7					-2	-2	-2	-2	-2	-1	-1	-1	-1	-1	0	0	0	0	0	0	0	0	0	0	0	0	0	0
8						-2	-2	-2	-2	-2	-2	-2	-1	-1	-1	-1	-1	0	0	0	0	0	0	0	0	0	0	0
9							-2	-2	-2	-2	-2	-2	-2	-2	-1	-1	-1	-1	-1	-1	0	0	0	0	0	0	0	0
10								-2	-2	-2	-2	-2	-2	-2	-2	-2	-1	-1	-1	-1	-1	-1	-1	0	0	0	0	0
11									-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-1	-1	-1	-1	-1	-1	-1	0	0
12										-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-1	-1	-1	-1	-1	-1	-1
13											-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-1	-1	-1	-1
14												-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-1	-1
15													-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
16														-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
17															-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
18																-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
19																	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
20																		-2	-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
21																			-2	-2	-2	-2	-2	-2	-2	-2	-2	-2
22																				-2	-2	-2	-2	-2	-2	-2	-2	-2
23																					-2	-2	-2	-2	-2	-2	-2	-2
24																						-2	-2	-2	-2	-2	-2	-2
25																							-2	-2	-2	-2	-2	-2
26																								-2	-2	-2	-2	-2

28 29

# Tstat as easy as mTPI to implement (spreadsheet computes table in previous slide)

Dose Selection Rules								
Value	Dose							
from	to	Increment						
-infinity	-2	2						
-2	-1	1						
-1	1	0						
1	2	-1						
2	-2							

