Dose-finding in oncology clinical trials based on unit probability mass

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Summary

- In oncology, patients are severely ill and often resort to advanced and aggressive treatments with potential serious toxicity
- The type of toxicity is called the dose-limiting toxicity (DLT), which could be as serious as permanent organ damage
- Oncologists want to find an appropriate dose level that is effective to the disease and yet is not "too toxic"
- **.** For example, < 30% of the patients will experience DLT
- The highest dose of which the probability of toxicity is less than p_T , say, $p_T = 0.30$, is called the maximum tolerated dose, or the MTD.

Fixed-dose setup:

- Align the doses from the lowest to the highest, say dose 1, 2, ..., 8.
- Usually starting from the lowest dose, although not necessary
- $\hfill \label{eq:linear}$ Treat the first cohort of patients (cohort size \geq 1) at the starting dose
- Depending on the observed toxicity from the treated patients, increase/decrease/not change the dose level for treating the next cohort

Variable-dose setup:

- the number of doses and their levels are not fixed
- the increment of the dose level depends on the type/grade of the toxicity

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Key criteria – SPST

Key points when developing or evaluating a design for phase I trials are SPST

- SAFE safety is the top priority and should be the first thing to check with any new design: scientific and legal implication
- Performs well in a variety of different scenarios a good design must perform reasonably well in various settings: simulation is the tool to examine the performance (more on this later).
- SIMPLE complicated designs are almost surely going to be declined by the physicians: they can ALWAYS use 3+3.
- Transparent it is easier for the physicians to understand if they can see how doses are assigned. They can even help to improve the design by pointing out unreasonable rules, e.g., escalate when 2 DLTs are found out of 4 patients.

Some common misconceptions

- Dose-escalation vs.estimation of the MTD
- Binary toxicity vs. graded toxicity
- Monotonicity is not a big deal
- Curve free vs. model free
- Tradeoff between percentage of correct selection vs. patients assignment

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The 3+3

Patients are enrolled in cohorts of 3. The core algorithm of the 3+3 contains:

- If none of the three patients in a cohort has a DLT, another cohort of three will be treated at the next higher dose.
- If one of the three patients has a DLT, three more will be treated at the same dose.
- The dose escalation continues until at least two patients of three or six patients have DLTs.

People add in additional rules (e.g., allowing for dose de-escalation

in the event of more than two DLTs) to make the algorithm more flexible.

Let us look at the SPST:

- Safe? YES. In general, the 3+3 is known to be conservative.
- Performs well? Not necessary, especially in cases with more doses.
- Simple? YES. No computation for trial conduct.
- Transparent? YES. The rules are laid out up front and clear.

So one of the "required" properties is not satisfied – does not always perform well.

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Evaluation of the 3+3 – con't

In addition:

- Problem number one: what if my $p_t = 0.1$? Can 3 patients estimate a 10% proportion?
- Problem number two: why would I want to stop at 6 patients regardless?
- Problem number three: A single DLT will forbid any further escalation. But DLT is a random event!

Example: if $p_i = 0.05$, $P(\ge 1 \text{ DLT in 3 patients}) = 0.14$. So even the true toxicity rate for a dose is 5%, one would at least stop escalatioin 14% of the time!

The first notable pulication about the CRM is O'Quigley, Pepe, and Fisher (1990) (Biometrics).

- Key idea: assuming a dose-toxicity response curve, the CRM continuously reassess the toxicity of all the doses and assign future patients to the dose closest to the MTD.
- Set the set of the
 - Need to specify the form of the dose response curve.
 - Need to specify a probability model with appropriate priors.
 - Need to calibrate the model parameters before trial starts so that appropriate operating characterisitics can be achieved.
 - Need to set up a web-interface to allow for real-time dose-assignment during trial conduct.

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The CRM – a specific model

Perhaps the most popular version of the CRM is the power model:

- The dose-response curve: $p_i = p_{i0}^{\exp(\alpha)}$, where p_{i0} are fixed known constants, and α is a parameter that describes the dose response curve.
- The prior for α is N(0,2).
- The p_{i0} 's are decided by solving $E[p_i^{\exp(\alpha)}] = s_i$, where s_i 's are a set of prior probabilities that one must determine (called "skeleton").

• The probability model is a binomial likelihood: $\prod_{i=1}^{d} p_i^{y_i} (1-p_i)^{n_i-y_i}$ where y_i and n_i are the number of DLTs and patients at dose *i*, respectively.

- Posterior of α is obtained by numerical integration.
- The next dose is $\arg \min_i |\hat{p}_i p_T|$, where \hat{p}_i is the posterior mean. BASS XVI, Savannah GA, 2009 - p. 10/45

The CRM – con't

- The dose-response curve: $p_i = p_{i0}^{\exp(\alpha)}$.
- The prior for α is N(0,2).
- Solving $E[p_i^{\exp(\alpha)}] = s_i$ for p_i (need to decide s_i).
- The next dose is $\arg \min_i |\hat{p}_i p_T|$.

One problem is to choose s_i . So far, one paper (to appear) claims it has an automatic way of choosing s_i . Most times, people do it arbitrarily. For example, with five doses, s_i 's are (0.1, 0.2, 0.3, 0.4, 0.5).

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The CRM – trial conduct

- The software only provides simulation results.
- How does one actually assign patients to doses when a trial starts?
- Answer:
 - Need to write another software involving numerical integration at real time.
 - Need to set up an interface between statisticians and nurses at clinics.
 - The interface allows the nurses to input the toxicity data, with which statisticians can compute the next dose to assign.

Let us look at the SPST.

- Safe? Somewhat, need additional rules.
- Performs well? Depending on the priors.
- Simple? NO.
- Transparent? No.

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A typical set of observation

Notation: <i>E</i> = Escalation; = Unacceptable (too toxic)	S = Stay; D.	= De-escalation;	U
Current dose	observed toxicity	Decision	
1	0/3	E	
2	0/3	E	
3	2/3	D	
2	2/6	S	
2			

At the end of the trial, one dose is selected as the estimated MTD. In the above case, probably dose 2 will be selected.

Toxicity probability interval method (TPI)



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Idea

Main idea:

- Suppose patients are currently being assigned to dose i, with toxicity probability p_i .
- Dose-assignment decisions are based on posterior toxicity probability that p_i is in one of the following three intervals: $\{(0, p_T K_1\sigma_i), [p_T K_1\sigma_i, p_T + K_2\sigma_i], (p_T + K_2\sigma_i, 1)\}$, where σ_i is the posterior standard deviation of p_i , and K_1 and K_2 are some small positive constants (default value $K_1 = 1.5, K_2 = 1$).
- If the posterior probability of the first, second, or the third interval is the largest, then the toxicity probability p_i is likely to be smaller to p_T , close to p_T , or larger than p_T , respectively, implying the dose level should be escalated, kept unchanged, or de-escalated.

- Likelihood function is a product of binomial densities: $l(\mathbf{p}) \propto \prod_{i=1}^{d} p_i^{x_i} (1-p_i)^{n_i-x_i}$, where n_i and x_i are the numbers of patients treated at dose i and experienced DLT, respectively.
- The priors of p_i are i.i.d. Beta (α, α) , where α takes a small value, e.g., $\alpha = 0.005$, resulting in a U-shaped prior.
- Posteriors are beta with known parameter values.

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Prior-posterior



Let D, S, E denote the decision to de-escalate to dose (i - 1), stay at dose i, and escalate to dose (i + 1), respectively. Following the main idea, define the posterior probabilities for the three intervals:

$$q(D, i) = P(p_i - p_T > K_1 \sigma_i | \text{data}),$$

$$q(S, i) = P(-K_2 \sigma_i \le p_i - p_T \le K_1 \sigma_i | \text{data}),$$

$$q(E, i) = P(p_i - p_T < -K_2 \sigma_i | \text{data}).$$

The dose-assignment rule

$$\mathcal{B}_i = \arg \max_{m \in \{D, S, E\}} q(m, i),$$

i.e. take the decision that has the maximum posterior probability.

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Two issues

- What if the first dose is very toxic?
- What if dose i 1 is safe, but dose i is very toxic?

Define

 $\mathcal{T}_i = 1\left\{ P(p_i > p_T | \mathsf{data}) > \xi \right\},\$

where 1{} is the indicator function and $\xi \in (0, 1)$ is a cutoff value (e.g., $\xi = 0.95$). For a large value of ξ , $T_i = 1$ implies that dose *i* is very likely to be highly toxic, and escalation to this dose should be permanently prohibited. To incorporate this rule, modified decision rule is given by

$$\mathcal{B}_i^{(e)} = \arg \max_{m \in \{D, S, \tilde{E}\}} q(m, i),$$

where $q(\tilde{E}, i) = q(E, i)(1 - \mathcal{T}_{i+1})$. Therefore, if $\mathcal{T}_{i+1} = 1$, the probability $q_{\tilde{E},i}$ equals zero and the assignment rule $\mathcal{B}_i^{(e)}$ can be only D, to deescalate, or S, to stay.

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Dose-finding algorithm

Suppose that the current tried dose is *i* for *i* = 1,...,*d*. After the toxicity outcomes of the last cohort are observed, select the dose for treating the next cohort among {(*i* − 1), *i*, (*i* + 1)} based on the assignment rule $\mathcal{B}_i^{(e)}$. There are two exceptions: if *i* = 1, the next available doses are {*i*, (*i* + 1)}; if *i* = *d*, the next available doses are {(*i* − 1), *i*}.

Dose-finding algorithm con't

- Suppose that dose 1 is a dose that has been tried previously. If $T_1 = 1$, terminate the trial due to excessive toxicity. Otherwise, terminate the trial when the maximum sample size is reached.
- In the special case of cohort of size 1, by convention, do not apply the exclusion rule T_i until two or more patients have been evaluated at a dose.

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Dose-finding algorithm con't

- At the end of the trial, select the dose as the estimated MTD with the smallest difference $|\hat{p}_i^* p_T|$ among all the tried doses *i* for which $T_i = 0$.
- Quantity \hat{p}_i^* is the isotonic transformation estimator of the posterior mean \hat{p}_i so that $\hat{p}_i^* \leq \hat{p}_i^*$ for j > i.
- If two or more doses tie for the smallest difference, perform the following rule. Let p* denote the transformed posterior mean of the tied doses.
 - If $p^* < p_T$, choose the highest dose among the tied doses.
 - If $p^* > p_T$, choose the lowest dose among the tied doses.

		Re	ecomme	endatio	n perce	entage	at dos	se leve	el		Tox	Ave
				$p_T =$	0.25						%	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 1		5	25	50	60	70	80	90	95	none		
Bayes	% MTD	13	79	8	0	0	0	0	0	0	25	30
	# Pts	7.7	16.1	5.8	0.5	0	0	0	0			
3+3	% MTD	24	58	16	2	0	0	0	0	0	25	12
	# Pts	4.0	5.0	2.6	0.4	0	0	0	0			
BCD	% MTD	10	78	11	1	0	0	0	0	0	24	30
	# Pts	11.4	11.5	5.2	1.4	0.3	0.1	0	0			
CFM	% MTD	6	80	14	0	0	0	0	0	0	29	30
	# Pts	5.2	16.3	7.5	0.9	0	0	0	0			
CRM	% MTD	6	83	11	0	0	0	0	0	0	27	30
	# Pts	5.7	18.6	4.9	1.0	0	0	0	0			

BASS XVI, Savannah GA, 2009 - p.25/45

		R	ecomn	nendat	ion pe	rcentag	ge at de	ose lev	vel		Tox	Ave.
				p_T :	= 0.25						%	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 2		1	2	3	4	5	6	25	50	none		
Bayes	% MTD	0	0	0	0	2	22	62	14	0	12	30
	# pt	3.2	3.2	3.4	3.5	3.7	4.5	5.9	2.6			
3+3	% MTD	0	0	0	2	3	21	46	8	0	11	27
	# pt	3.1	3.2	3.3	3.4	3.3	3.7	4.5	2.2			
BCD	% MTD	0	0	1	2	7	24	56	10	0	10	30
	# pt	3.2	3.6	3.6	3.5	3.8	5.4	4.8	2.1			
CFM	% MTD	0	0	0	0	1	22	61	16	0	12	30
	# pt	3.1	3.0	3.1	3.5	3.7	5.1	6.3	2.1			
CRM	% MTD	0	0	1	1	5	22	50	21	0	13	30
	# pt	3.1	3.4	3.3	3.7	3.6	4.4	5.1	3.4			

		R	lecomm	endatior	n perce	entage	at dos	se leve	el		Тох	Ave.
				$p_T =$	0.25				_		%	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 3		1	5	50	60	70	80	90	95	none		
Bayes	% MTD	0	79	21	0	0	0	0	0	0	22	30
	# pt	5.5	13.2	10.2	1.0	0	0	0	0			
3+3	% MTD	0	70	28	2	0	0	0	0	0	22	13
	# pt	3.1	5.2	4.4	0.7	0.1	0	0	0			
BCD	% MTD	0	60	39	1	0	0	0	0	0	22	30
	# pt	4.9	14.3	8.2	2.2	0.4	0	0	0			
CFM	% MTD	0	56	44	0	0	0	0	0	0	28	30
	# pt	3.1	11.7	13.1	2.0	0.1	0	0	0			
CRM	% MTD	0	49	51	0	0	0	0	0	0	26	30
	# pt	3.1	13.0	12.0	1.8	0	0	0	0			

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		Red	comme	ndatio	n perc	entage	e at do	se lev	el		Tox	Ave
				$p_T =$	0.25						%	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 4**		40	50	60	70	80	90	95	99	none		
Bayes	% MTD	31	2	0	0	0	0	0	0	67	41	19
	# pt	16.8	1.8	0.2	0	0	0	0	0			
3+3	% MTD	38	9	1	0	0	0	0	0	52	43	6
	# pt	4.7	0.5	0.6	0.7	0	0	0	0			
BCD	% MTD	38	2	0	0	0	0	0	0	60	45	18
	# pt	12.6	4.6	1.2	0.2	0	0	0	0			
CFM	% MTD	38	3	1	0	0	0	0	0	58	42	14
	# pt	11.7	1.9	0.5	0.1	0	0	0	0			
CRM	% MTD	47	2	0	0	0	0	0	0	51	42	23
	# pt	20.2	2.5	0.2	0	0	0	0	0			

		Re	ecomme	ndatio	n perce	entage	at dos	se leve	el		Тох	Ave.
				$p_T =$	0.25				_		%	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 5		15	25	35	45	55	65	75	85	none		
Bayes	% MTD	31	41	21	7	0	0	0	0	0	24	30
	# pt	12.4	9.5	5.5	1.9	0.3	0	0	0			
3+3	% MTD	29	37	20	7	1	0	0	0	8	26	12
	# pt	4.4	3.9	2.4	0.9	0.2	0	0	0			
BCD	% MTD	21	46	22	6	1	0	0	0	5	26	29
	# pt	10.6	9.2	5.7	2.5	0.8	0.1	0	0			
CFM	% MTD	15	44	32	7	0	0	0	0	0	27	30
	# pt	8.0	10.6	8.0	2.6	0.4	0	0	0			
CRM	% MTD	36	47	14	2	0	0	0	0	0	23	30
	# pt	13.8	11.4	3.6	0.9	0.2	0	0	0			

BASS XVI, Savannah GA, 2009 - p.29/45

		F	Recomm	nendatio	n perc	entage	at dos	se leve	I		Tox	Ave.
				$p_T =$	0.25						%	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 6		5	15	25	35	45	55	65	75	none		
Bayes	% MTD	2	24	42	24	7	0	0	0	0	22	30
	# pt	5.1	8.2	9.2	5.7	1.6	0.3	0	0			
3+3	% MTD	9	28	34	22	5	0	0	0	0	21	15
	# pt	3.6	4.3	3.8	2.3	0.8	0.2	0	0			
BCD	% MTD	1	29	44	19	6	1	0	0	0	21	30
	# pt	6.8	8.7	7.5	4.4	1.9	0.6	0.1	0			
CFM	% MTD	0	14	49	29	6	0	0	0	0	24	30
	# pt	3.9	6.2	10.7	7.1	1.8	0.3	0	0			
CRM	% MTD	4	37	45	12	2	0	0	0	0	20	30
	# pt	5.5	11.5	8.9	3.4	0.7	0.1	0	0			

Evaluation of the TPI

Let us look at the SSTP:

- Simple? YES.
- Safe? YES.
- Transparent? YES. Ĵ
- Performs well? YES.

Looks like we have got a perfect design. But...

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TPI – Not there yet

Two issues:

- The interval rule is based on $(p_T K_1\sigma_i, p_T + K_2\sigma_i)$.
 - σ_i is the posterior standard deviation of p_i easy to compute (in closed form).
 - K_1 and K_2 are fixed and must be given. Default values are $K_1 = 1.5$ and $K_2 = 1$. However, they need to be calibrated for certain trials.

The number of patients treated at doses above the MTD is a bit higher (compared to the CRM, for example) - room for improvement.

The modified TPI (mTPI) method

The mTPI is based on a new statistics called the unit probability mass. The mTPI improves the TPI on two aspects:

- mTPI is calibration free does not require tuning of parameters
- mTPI is safer treats fewer patients at over-toxic doses while maintaining other good performance properties of the TPI.

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The mTPI – introduction

For interval (a, b), the UPM is defined as

$$\frac{F(b) - F(a)}{b - a},$$

where $F(\cdot)$ is a probability distribution function. In the mTPI method, the framework for dose-findin is the following:

- Define an equivalence interval (EI) $I_{EI} = (p_T \epsilon_1, p_T + \epsilon_2)$, where any dose with probability of toxicity within the interval can be considered as a MTD.
- Compute the UPM of three intervals, $I_{safe} = (0, p_T \epsilon_1)$, $I_{toxic} = (p_T + \epsilon_2, 1)$ and I_{EI} under the posterior distribution of p_i .
- Choose to escalate, de-escalate, or stay at dose *i* if the UPM of I_{safe} , I_{toxic} , or I_{EI} is the largest, respectively.



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The mTPI – EI

The equivalence interval $I_{EI} = (p_T - \epsilon_1, p_T + \epsilon_2)$ is the only thing that one needs to decide.

- We will show that the choices of ϵ_1 and ϵ_2 do not affect the performance of the mTPI.
- We should elicit the EI as a general practice: since we never get a dose with exact toxicity probability p_T .

		F	Recomm	endati	on per	centag	e at do	ose lev	əl		Тох	Ave
				$p_T =$	= 0.25						pctg*	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 1		5	25	50	60	70	80	90	95	none		
mTPI	% MTD	14	78	8	0	0	0	0	0	0	24	30
	# Pts	7.1	18.3	4.4	0.2	0	0	0	0			
TPI	% MTD	13	79	8	0	0	0	0	0	0	25	30
	# Pts	7.7	16.1	5.8	0.5	0	0	0	0			
CRM	% MTD	6	83	11	0	0	0	0	0	0	27	30
	# Pts	5.7	18.6	4.9	1.0	0	0	0	0			
Scenario 2		1	2	3	4	5	25	50	60	none		
mTPI	% MTD	0	0	0	2	16	71	10	1	0	16	30
	# Pts	3.2	3.5	3.5	4.0	5.2	8.1	2.3	0.1			
TPI	% MTD	0	0	0	0	19	70	11	0	0	15	30
	# pt	3.2	3.2	3.3	3.6	5.0	8.0	3.3	0.3			
CRM	% MTD	0	0	1	1	20	61	16	2	0	16	30
	# pt	3.1	3.4	3.3	3.7	4.7	7.0	3.8	0.9	BASS	XVI, Savannah (GA, 2009 – p.37

		Re	ecomme	ndation	perce	ntage	at dos	e leve	el		Tox	Ave
				$p_T = 0$	0.25						pctg*	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 3		1	5	50	60	70	80	90	95	none		
mTPI	% MTD	0	82	17	0	0	0	0	0	0	21	30
	# pt	3.2	15.9	10.3	0.6	0	0	0	0			
TPI	% MTD	0	79	21	0	0	0	0	0	0	22	30
	# pt	5.5	13.2	10.2	1.0	0	0	0	0			
CRM	% MTD	0	49	51	0	0	0	0	0	0	26	30
	# pt	3.1	13.0	12.0	1.8	0	0	0	0			
Scenario 4		40	50	60	70	80	90	95	99	none		
mTPI	% MTD	31	2	0	0	0	0	0	0	67	41	19
	# pt	16.8	2.0	0.2	0	0	0	0	0			
TPI	% MTD	31	2	0	0	0	0	0	0	67	41	19
	# pt	16.8	1.8	0.2	0	0	0	0	0			
CRM	% MTD	47	2	0	0	0	0	0	0	51	42	23
	# pt	20.2	2.5	0.2	0	0	0	0	0	BASS	XVI, Savannah (GA, 2009 –

		Re	ecomme	endatio	n perc	entage	at dos	se leve	el		Tox	Ave
				$p_T =$	0.25						pctg*	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 5		15	25	35	45	55	65	75	85	none		
mTPI	% MTD	29	45	20	4	0	0	0	0	0	24	30
	# pt	12.4	10.9	5.0	1.1	0.1	0	0	0			
TPI	% MTD	31	41	21	7	0	0	0	0	0	24	30
	# pt	12.4	9.5	5.5	1.9	0.3	0	0	0			
CRM	% MTD	36	47	14	2	0	0	0	0	0	24	30
	# pt	13.8	11.4	3.6	0.9	0.2	0	0	0			
Scenario 6		5	15	25	35	45	55	65	75	none		
mTPI	% MTD	2	28	42	23	4	0	0	0	0	20	30
	# pt	4.9	10.2	9.3	4.5	0.9	0.1	0	0			
TPI	% MTD	2	24	42	24	7	0	0	0	0	22	30
	# pt	5.1	8.2	9.2	5.7	1.6	0.3	0	0			
CRM	% MTD	4	37	45	12	2	0	0	0	0	20	30
	# pt	5.5	11.5	8.9	3.4	0.7	0.1	0	0	BASS	XVI, Savannah (GA, 2009 – p.3

Sensitivity to the El

		Re	ecomme	endatio	n perc	entage	e at do	ose lev	/el		Тох	Ave
				$p_T =$	0.25						pctg*	n
	Dose	1	2	3	4	5	6	7	8			
Scenario 1		5	25	50	60	70	80	90	95	none		
$\epsilon_1 = \epsilon_2 = .05$	% MTD	14	78	8	0	0	0	0	0	0	24	30
	# Pts	7.1	18.3	4.4	0.2	0	0	0	0			
$\epsilon_1 = \epsilon_2 = .2$	% MTD	15	76	9	0	0	0	0	0	0	24	30
	# Pts	7.7	18.4	3.7	0.2	0	0	0	0			
$\epsilon_1 = \epsilon_2 = .001$	% MTD	14	78	8	0	0	0	0	0	0	24	30
	# Pts	7.1	18.3	4.3	0.2	0	0	0	0			

The mTPI and TPI are SSTP.

- The mTPI is safer.
- The mTPI is calibration free.

I would now only use the mTPI. So do we have the same Excel software?

BASS XVI, Savannah GA, 2009 - p.41/45

Software

Yes. Both mTPI amd TPI are available to download for free.

Paper and software http://odin.mdacc.tmc.edu/~yuanj/

Simulation-based evaluation

The choice of scenarios is rarely discussed in the literature. However, it is critical as failure to include important scenarios may lead to undesirable consequences in practice.

- Early-stopping scenarios (0.50, 0.60, ...)
- Fast-escalating scenarios (0.01, 0.03, 0.05, 0.10, 0.15, 0.3)
- Big-gap scenarios (0.05, 0.10, 0.50, 0.60, ...)
- Regular scenarios (0.10, 0.20, 0.30, 0.40, ...)

BASS XVI, Savannah GA, 2009 - p.43/45

Happy dose-finding!

Thank you!