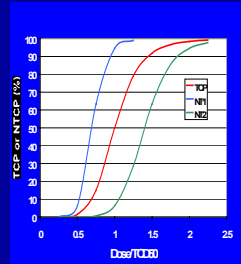


Biological Indices for IMRT Evaluation and Optimization

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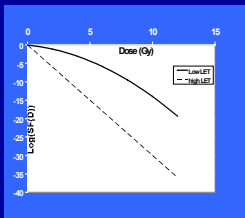
Why biological indices?



- Dose distributions, DVHs-surrogates for outcome
- We want:
 - high local control (TCP~100%)
 - low complications (NTCP~0% for all tissues)
- Wanted! reliable mathematical models for
 - TCP (D(t),V)
 - NTCP(D(t),V)

Linear-Quadratic (LQ) model

SF(D)= mean fraction of cells surviving dose D



Log₁₀ SF(D) vs Dose [single treatment]

$$SF(D) \sim \exp(-[\alpha D + \beta D^2])$$

- α Single track, non-repairable, lethal damage
- β '2-track', partially repairable, sublethal damage

For low LET

- α : 0.1 Gy⁻¹ – 1.5 Gy⁻¹
- α/β : 1.0 Gy – 20 Gy
- α hard to measure vs α/β

SF(2) or SF2=fraction of cells surviving a single 2 Gy dose
Back-of-envelope estimates
SF2~0.5

Fractionated Doses

$$SF(n,d) = (\exp(-[\alpha d + \beta d^2]))^n$$

$$SF(D,d) = \exp(-\alpha D [1 + d/(\alpha/\beta)])$$

where n=# of fractions, d=dose/fx, D=nd

- Large α/β (~10 Gy)->“Early responding” tissues
 - Insensitive to dose per fraction
 - most tumors, mucosa, skin- acute complications
- Small α/β (<~4 Gy)->“Late responding” tissues
 - Sensitive to dose per fraction
 - spinal cord, lung, kidney – late complications
 - maybe prostate tumors (α/β estimates from 1 Gy-10 Gy)
 - other slow growing tumors

Biological Effective Dose (BED)

Schedules with same BED have same biological effect

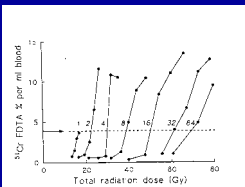


Figure 13.3 Dose-response curves for late damage to the mouse kidney with fractionated radiation exposure. Damage is indicated by EDTA clearance, curves determined for 1 to 64 dose fractions, illustrating the sparing effect of treatment fractionation. From Stewart et al (1984), with permission.

$$BED = D(1 + d/(\alpha/\beta))$$

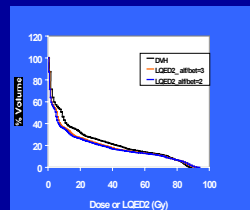
Tumor $\alpha/\beta=10$ Gy
Normal tissue $\alpha/\beta=3$ Gy
Standard schedule: D=60, d=2 Gy
 $BED_{tum,0}=72$ Gy $BED_{NT,0}=100$ Gy

Hyperfractionation

At 1.2 Gy/fx
Iso-tumor-effect-> $BED_{tum}=72$ Gy
Dose= $72/1.12 \sim 64.3$ Gy ->64.8 Gy
 $BED_{NT}=90.7$ Gy -> less NT damage
Iso-NT-effect-> $BED_{NT}=100$ Gy
Dose= $100/1.4 = 71.4$ Gy->70.8 Gy
 $BED_{tum}=79.3$ Gy->higher tumor effect
Either way, better therapeutic ratio

LQED2 (or ED2)

Dose given at 2 Gy/fx with same bio-effect as (D,d)



$$LQED2 = \frac{D(1 + d/(\alpha/\beta))}{(1 + 2/(\alpha/\beta))}$$

Transform each DVH dose bin separately.

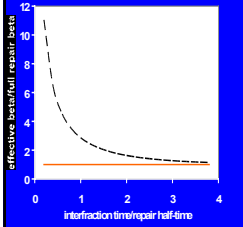
LQED2 <total dose if d< 2Gy/fx
LQED2 >total dose if d>2Gy/fx

Lung DVH - 80 Gy treatment (2 Gy/fx to 100% isodose) and LQED2-VH's
 $\alpha/\beta=3$ Gy (red)
 $\alpha/\beta=2$ Gy (blue)

Time dependence: Sublethal Damage Repair

“Treat at least 6 HOURS APART”

Occurs for tumors & normal tissue
Initial evidence - “split dose experiments”



Effective β vs ΔT (30 fractions)

At constant (d, n)
Short inter fraction time ΔT
Less time for repair, higher β

Sublethal damage kinetics ~
 $\exp(-\text{Ln}2 \Delta T/T_{1/2})$
 $T_{1/2}$ range ~0.3 hr to 4 hr
 $T_{1/2}$ depends on the tissue

$$\beta(\Delta T) = \beta(1 + h(n, \Delta T))$$

(particular concern for LDR)

Time Dependence: Repopulation

Cells proliferate during a course of treatment

Long RT course \rightarrow more time to proliferate, lower tumor control

Growth ~ exponential over treatment course T
 $SF(n, d, T) \sim \exp[-\alpha nd(1 + d/\{\alpha/\beta\})] \exp(\text{Ln}2 T/T_{\text{pot}})$
 T_{pot} range 2 days – 100 days

“Back-of envelope” BED lost to tumor growth ~ 0.5 Gy/day

Continuous Hyperfractionated Accelerated RT (CHART)
Control (H&N): 2 Gy/fx to 66 Gy, 1 Tx/dy, weekends off, T=45 dy
CHART: 1.5 Gy/fx to 54 Gy, 3 Tx/dy, $\Delta T > 6$ hr, T=12 dys continuous
CHART expected to gain ~ 16 Gy of BED from reduced T
Clinical results: CHART “at least as effective” as control, trend shows benefit for advanced disease.

LQ Model Summary

- Grounded in decades of in vitro, in vivo work
- Widely used to compare, devise Tx schedules
- Can include time-dependent effects
- Key parameters have large uncertainties, even for “common” systems (e.g. prostate cancer)
- Mechanistic understanding imperfect
- Not strictly “rigorous”

BUT it works and is clinically very useful

TCP: Modeling problems

- TCP endpoint = local control *but* at what time?
- Where is the tumor?
 - Tumor localization on planning images?
 - Multi-modality imaging?
 - “Local failure” or “marginal miss”?
- Which tumor cells determine local control?
- What was the delivered dose distribution?
 - Role of setup errors, organ motion?
- Uncertainty in radiobiological parameters.

Tumor Control Probability (TCP)

TCP = “long-term” local control
~ Sigmoidal with dose (??)

D50 = local control dose for 50% of cases (also called TCD50)

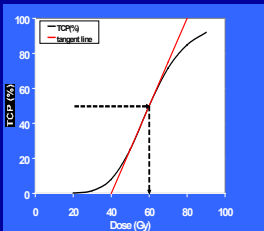
γ_{50} proportional to slope of TCP vs dose at D50

$$\gamma_{50} = \frac{\Delta \text{TCP}(\%)}{\Delta(100 D/D50)}$$

Parameter Range (Std Fx)

D50: 20 Gy to >100 Gy

γ_{50} : 1-4



Example TCP curve
D50=60 Gy, $\gamma_{50}=1.5$

TCP: Poisson Model

Only clonogenic cells can regrow the tumor

TCP = probability of no surviving clonogens

Let N = # clonogens, SF(D) = mean fraction clonogens surviving D
(N SF(D)) = mean # of surviving clonogens
Poisson distribution of surviving clonogens

$$\text{TCP} \sim \exp(-N \text{SF}(D))$$

compatible with models of SF such as LQ

$$\text{TCP} \sim \exp(-N \text{SF}^2(D)/2)$$

$$\text{TCP}_{\text{LQ}} \sim \exp(-N \exp[-\alpha D(1 + d/\{\alpha/\beta\})])$$

D50 ~ $2 \text{Ln}(\text{Ln}2/N) / \text{Ln}(\text{SF}2)$

$\gamma_{50} = \text{Ln}2/2 \text{Ln}(N/\text{Ln}2)$ independent of SF2 or (α, β)

Problems, problems.....

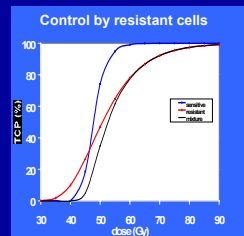
Soft tissue cell density $\sim 10^9$ cells/cc
 Detectable tumors >1 cc (100 cc tumors are common)
 If $N \sim \# \text{cells}$, $N \geq 10^9 \rightarrow \gamma_{50} \geq 7.3$
 This contradicts clinical observation: $\gamma_{50} \sim 2$

Two possible solutions

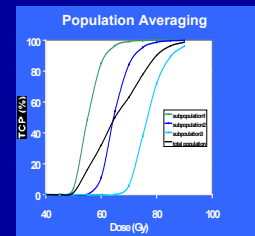
1. TCP controlled by a few radioresistant clonogens ($N \sim 200 \rightarrow \gamma_{50} \sim 2$)
2. TCP is a population average:
 (inter-tumor) different tumors have different radiosensitivities
 (intra-tumor) clonogens within a tumor vary in radiosensitivity.

These are not mutually exclusive

Two ways to reduce γ_{50}



Blue: 10^7 sensitive clonogens, $SF_2=0.5$
 Red: 200 resistant clonogens, $SF_2=0.8$
 Black: tumor's clonogens a mixture



Patient population is 3 equal sized groups w different radiosensitivities-
 Observed TCP=population average

TCP for non-uniform dose distributions

Divide tumor into subvolumes (tumorlets)

i^{th} tumorlet: volume= v_i , # & density of clonogens= n_i, ρ_i
 Dose and dose/fx D_i, d_i ; surviving fraction $SF(D_i, d_i)$

TCP=probability of no survivors in any tumorlet

Assume Poisson model, independence of each tumorlet

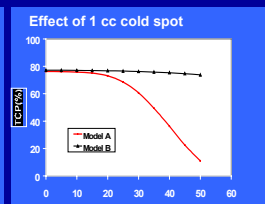
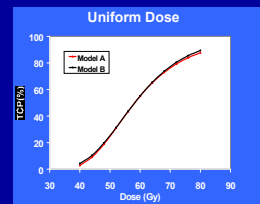
$$TCP = \prod \exp(-n_i SF(D_i, d_i)) = \prod \exp(-\rho_i v_i SF(D_i, d_i))$$

Product over all the tumorlets

Geographic misses and deep cold spots are bad!
 $D \rightarrow 0$ in a tumorlet $\rightarrow TCP \sim 0$

Effect of cold spot has implications for use of TCP in plan optimization but.....

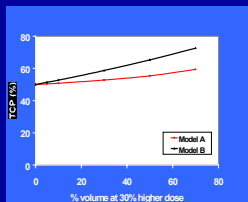
Effect of cold spot is model dependent



Model A: Population avg model, Webb parameters, 10^7 clonogens/cc, 100 cc
Model B: 90 radioresistant clonogens control TCP

99% of the tumor is at 70 Gy
 X-axis =% underdose to 1% (1 cc)
 Uniform clonogen density
 Low clonogen # \rightarrow insensitive to cold spot.

Small hotspots don't help much A model-robust conclusion



Same models as previous slide.
 Uniform clonogen density. Varying percent volume (x axis) receives 30% extra dose, remaining tumor at D_{50} .

Small hot spots don't improve TCP much UNLESS they coincide with small regions of radioresistant or a high density of clonogens ("dose painting")

Big hot spots=Boosts

EUD:Equivalent Uniform Dose

EUD=uniform dose giving same clonogen survival as the true dose distribution (Niemierko, 1997)

For large α/β and uniform clonogen density

$$EUD = 2 \ln(\sum v_i (SF_2)^{D_i/2}) / \ln(SF_2)$$

 v_i =volume fraction receiving dose D_i

EUD is independent of # of clonogens
 EUD is less model dependent than TCP.

$$D_{\min} \leq EUD \leq D_{\text{mean}}$$

Generalized for spatial clonogen variation, general α/β , LQ+time dependence, population averaging

TCP Summary

Poisson statistics +LQ Model mathematically tractable
Applicable to general dose distributions
BUT

There are major conceptual questions

How many clonogens are there? How many control TCP? What LQ parameters? Importance of population averaging?

Spatial distribution of clonogens in tumor is unknown

If a cold/hot spot has no clonogens, TCP or EUD isn't affected.

We assume uniform clonogen density out of ignorance.

Is there hope from molecular imaging?

Affect on TCP predictions for evaluation & optimization ??

Poisson model breaks down for proliferating cells (worse for short T_{pot})
impact on optimization/evaluation unknown

(see Stavrev et al, Med Phys May 2003)

Problems with modeling NTCP (Normal tissue complication probability)

- Reported clinical data mostly for low NTCP and low (<70 Gy) doses.
 - If NTCP increases sigmoidally with dose, most data are on the early "tail"- extrapolation is troublesome
- Several types of complications per organ
 - Different onset times, dose-volume dependences
- Reports use different scoring systems
- Non-radiation factors affect NTCP
- Most models quantize the complication
 - Most complications show severity continuum

Main data source – Emami et al, 1991

IJROBP 21, 109-122, 1991

- Critical review and summary of dose limiting complications literature up to 1991 for 28 normal tissues
- Predates 3D-CRT era (*time for new look?*)
- Tolerance doses: TD50/5 and TD5/5
 - Doses for complication probability of 50% and 5% respectively at 5 years, tx at 1.8-2 Gy/Fx

Volume dependence of tolerance doses

Dose=0

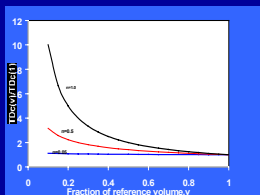
Dose=D
Volume fraction=v

Partial organ irradiation
Volume fraction v gets full dose
Remainder of organ gets zero dose

Emami et al tabulated TD50/5 and TD5/5 for partial organ irradiations of v=1, 2/3 and 1/3.

NTCP Volume Dependence

Burman et al, IJROBP 21, 123-135, 1991



TD for NTCP of c% vs volume fraction

Low n->spinal cord, brainstem
High n->lung (pneumonitis), parotids (xerostomia), liver (RILD)
Mid n-> rectum (bleeding), heart

- For most normal tissues iso-complication dose increases if irradiated volume fraction decreases
- Power law approximation

$$TDc(v) = TDc(1)v^n$$

$$n \geq 0$$

Small n->weak volume depdce, D_{max} dominates

High n->strong volume depdce (n=1 ->mean dose depdce)

Normal Tissue Complications Spinal Cord

- Complication=Radiation myelitis
- Clinically-want NTCP<<5%
- 1991(E&B) TD5~47-50 Gy, TD50~70Gy, n=0.05

Updates

- TD5~57 Gy
- Weak volume dependence confirmed
- Evidence for slow repair component (>8 hr)
- Small α/β (~2 Gy)
- Some occult injury recovery (~ 2 yrs)

Normal Tissue Complications Lung

- Complication=severe radiation pneumonitis
 - Requiring serious medical care
- NTCP~20-25% (steroids) accepted
- Onset within 6 months of Tx
- 1991 TD50(1)=24.5 Gy, TD5(1)=17.5 Gy, n=0.87

Updates

- n~1, TD50(1)~28 Gy
- Good DVH correlates for treatment planning:
 - D_{mean} , Volumes receiving >13 Gy, >20 Gy, >30 Gy
- α/β ~ 2-4 Gy
- Are some subvolumes more sensitive than others??

Normal Tissue Complications Late rectal complications

- Complications=severe proctitis, rectal bleeding, ulceration, stricture, fistula
- Onset latency up to 2-3 yrs post treatment
- NTCP~10-20% (bleeding) are accepted
- 1991 TD50=80 Gy, TD5=60 Gy, n=0.12

Updates

- α/β ~3-4 Gy
- More complex volume dependence (not just D_{max})
- MSK: Restrict volume >75 Gy, volume >~50 Gy
- DV constraints- "Planning" rectal wall 1 slice above and below PTV; <30% to >75.6 Gy, <53% to >47 Gy; D_{max} <84-86 Gy*

NTCP Models: Lyman Model

4-parameter sigmoidal function for all complications

$$NTCP = \int_0^{\infty} \exp(-t^2/2) dt / \sqrt{2\pi}$$

Parameters: TD50(1), m, volume exponent n, $V_{reference}$

Histogram reduction schemes

DVH ->equivalent partial organ or uniform irradiation

KB: use equivalent uniform dose $D_{eff} = (\sum v_i (D_i)^{1/n})^n$

Lyman: use equivalent partial irradiation of v_{eff} to D_{max}

$$v_{eff} = \sum v_i (D_i/D_{max})^{1/n}$$

Tissue Architecture Models

Tissue made of independent functional subunits (FSUs)
NTCP depends on FSU radiosensitivity and organization
Unfortunately FSUs remain speculative and un-identified!

Serial: X X X X X and X X X X X give same effect

NTCP=probability of no FSUs destroyed

Mathematically tractable (FSU response directly from whole organ)

Low volume dependence, applied to spinal cord

Parallel: Complication only if > a critical % of FSUs destroyed

Applied to strong volume dependence organs (lung, liver, kidney)

Need population averaging (over critical %) to reduce γ_{50}

NTCP saturates for high dose partial organ irradiation

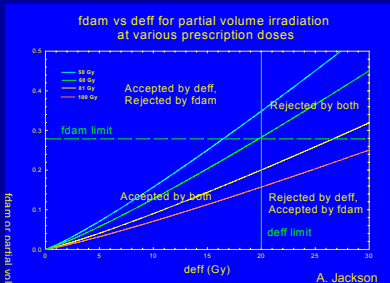
(e.g. partial irradiation of 25% of organ volume -> NTCP ≤ 25%)

The model can affect clinical decisions

An example for lung treatment plans with outcomes-based criteria.

f_{dam} =fraction damaged FSUs in parallel model

The Lyman model criterion ($d_{eff} < 20$ Gy) rejects some distributions that are accepted by the parallel model ($f_{dam} < 0.28$) and vice versa.



Generalized EUD for normal tissues

Calculated from the DVH, $\{v_i, D_i\}$

$$EUD = (\sum v_i (D_i)^a)^{1/a} \quad (\text{Niemierko, 1999})$$

EUD=Lyman model+KB histogram reduction D_{eff} , $a=1/n$

Differs from original EUD in that it is not mechanistic

Advantages

Computationally simpler than model NTCPs

Only one parameter/complication

Same formalism for tumors and normal tissues

{ a =large negative # for tumors so EUD is determined by the cold spot}

NTCP: Summary

Bad news

NTCP models are much too simplistic to describe the physiology of radiation damage. Maybe we shouldn't try!

Good news-maybe we don't need very sophisticated models
Crudely, there are 3 types of normal tissues

"Max dose" [serial?] tissues (cord, optic structures, bowel)

"Mean dose" [parallel??] tissues (lung, liver, parotids)

Mixed tissues (rectum) – look at middle parts of DVH

Such surrogates for NTCP are easily obtained from DVH
They are used in clinical decisions and IMRT optimization.
(mean lung dose < 21 Gy, Max cord dose < 50 Gy, < 30% RW over 75 Gy)
More good news: Lots of data from modern clinical studies!

Biological models in optimization score functions

Common features

- All the uncertainties associated with biological models
- Is the whole PTV equally important for TCP?
- False normal structure to limit hotspots in target
- More computation-intensive than dose, dose-volume

1. Uncomplicated Control [Agren et al, 1990]

Probability of local control (benefit) without injury

$$P_+ = P_B - P_I + \delta P_I (1 - P_B)$$

$P_B = \text{TCP}$, $P_I = 1 - \prod_k (1 - \text{NTCP}_k)$, $\delta = \text{independent B and I}$

$$\delta = 1 \rightarrow P_+ = P_B (1 - P_I)$$

Problem: does not discriminate against serious complications (5% TCP decrease same as 5% myelitis increase)

2. TCP and NTCP with prioritization

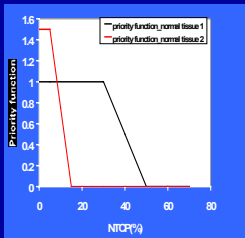
Maximize F where

$$F = \text{TCP} \prod_{NT} S_{NTi}$$

S_{NTi} is constant until
NTCP of ith organ hits
first threshold; falls to
zero at second threshold

Any TCP, NTCP model.

Results were compared with
dose but not dose-volume
constraints [Wang et al, 1995]



3. Generalized EUD : $(\sum v_i (D_i)^a)^{1/a}$

EUD constraint values set for each structure

(e.g. for tumor, EUD=the prescription dose)

Generalized EUD for each tissue of interest ($a < 0$ for tumors) calculated in each optimization iteration.

Quadratic score function (difference of calculated and constraint EUDs rather than doses) [Jones & Hoban, 2001]

Product of logistics score function [Wu et al, 2001]

$$S = [1 / (1 + (\text{EUD}_{0,\text{tum}} / \text{EUD}_{\text{tum}})^n)] \prod_{NT} [1 / (1 + (\text{EUD}_{NT} / \text{EUD}_{0,NT})^n)]$$

Better than dose/dose volume methods?

Wu et al find advantages:

Better normal tissue protection (protection throughout optimization, larger search space explored).

Only one constraint per tissue (vs several for dose, dose-volume constraints).

In a later report, cited need to combine EUD with dose-based score function to "fine tune" the plan for ultimate clinical use [Wu et al PMB 48, 2003]

Jones & Hoban: EUD gave no advantage.

Many questions remain.....

Do objective functions using biological models give better or more efficient IMRT distributions than dose/dose-volume functions?

How to do a fair comparison?

Plan evaluation with models or D-V correlates or both? How realistic do models need to be? Can we beat biologists at their own game (should we be trying)?

If multiple normal tissues compete with tumor control, how to weight models vs "clinical judgement" factors?

