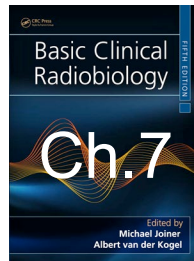
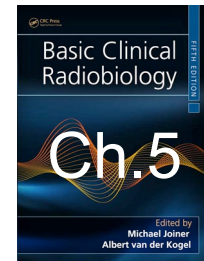


Basic Clinical Radiobiology

Dose-response relationships in radiotherapy

Michael Joiner

Toronto 2023

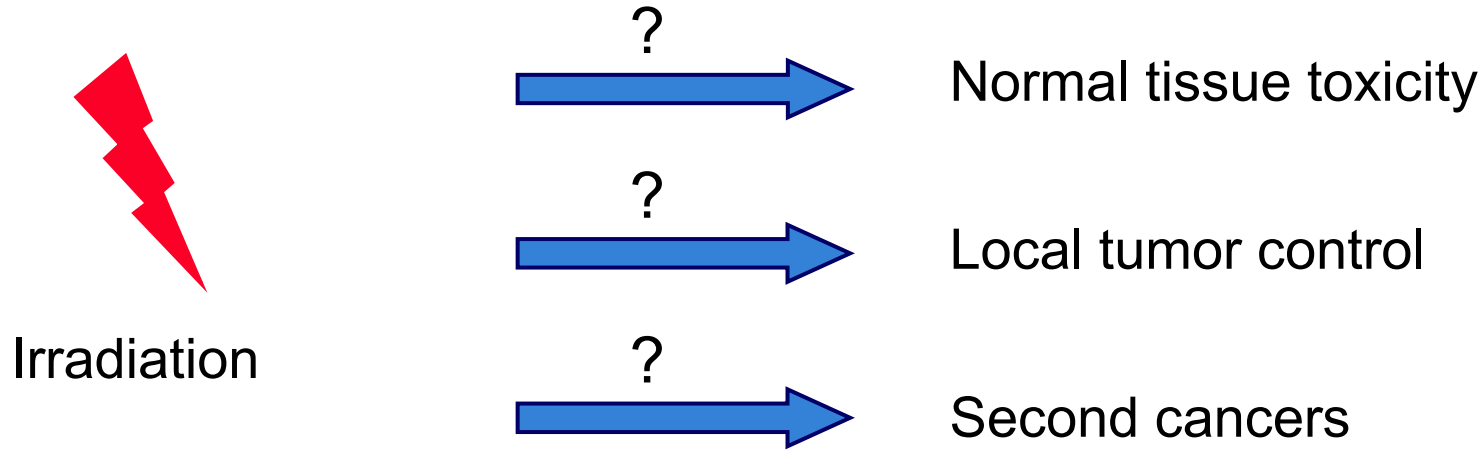


Definitions

Dose Response: Relationship between a given physical absorbed dose and the resulting biological response

Endpoint: A specific event that may or may not have occurred at a given time after irradiation

Dose responses – always document!

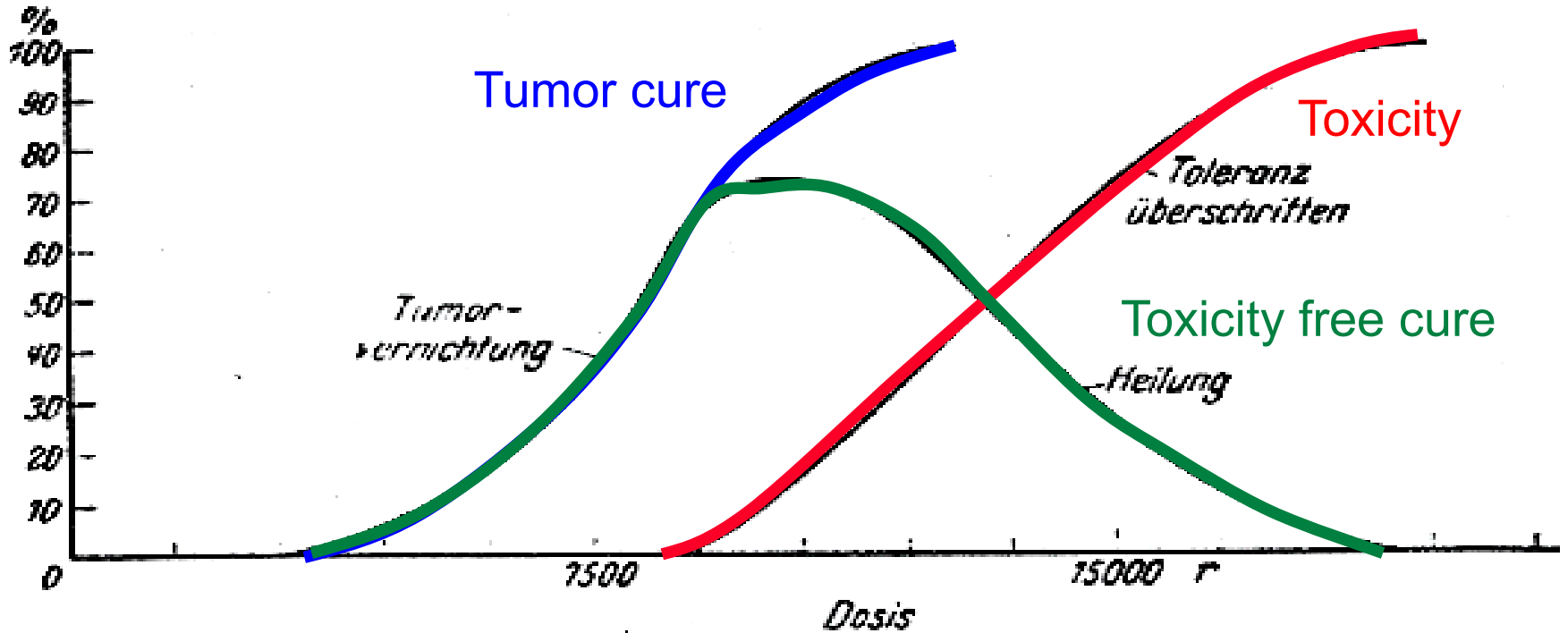


Relationship between given dose and each clinically relevant outcome needs to be defined

i.e. Define the incidence or probability of a certain outcome after a defined dose

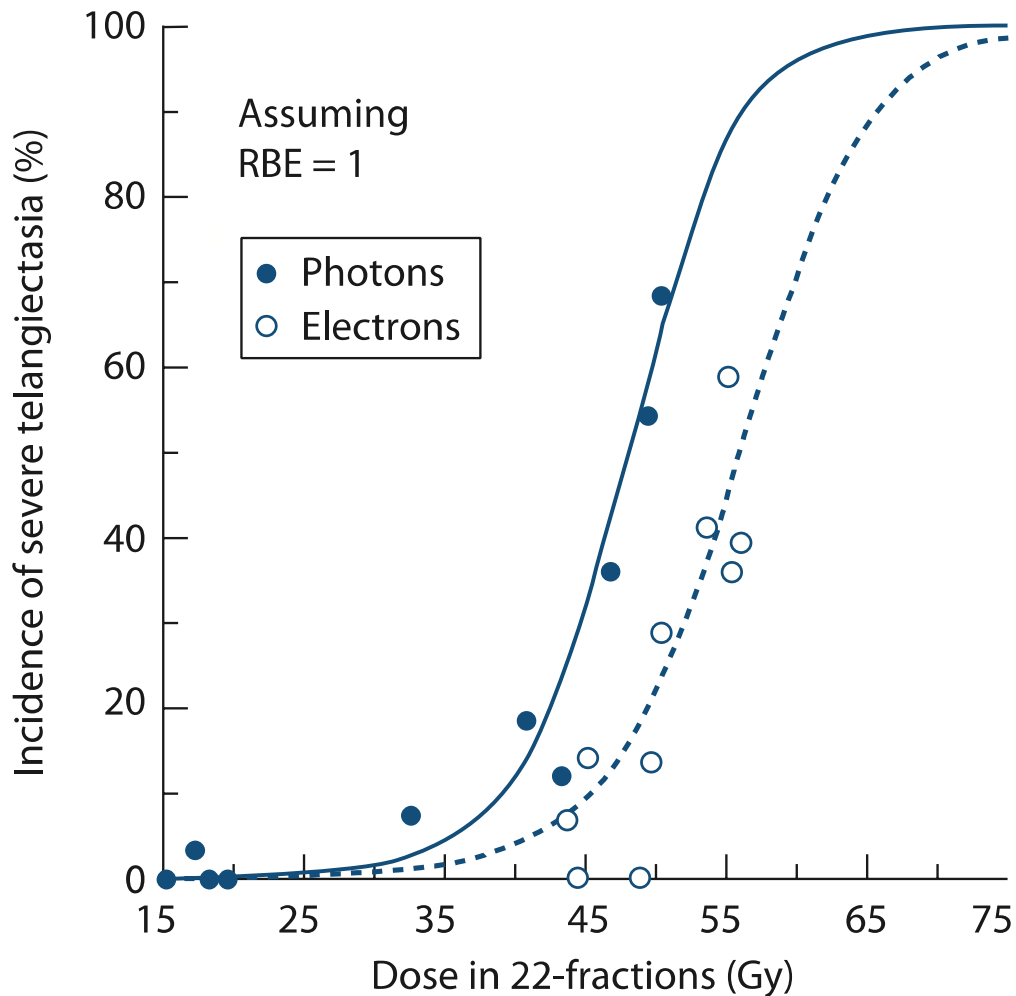
Dose response: Empirical data

Sigmoid curves indicate variability of clinical radioresponse



Examples of dose response relationships

Sigmoid curves



Dose response models

Most frequently used models to fit sigmoid dose-response curves:

- Poisson model ...tumor
- Logistic model ...normal

Dose response model: Tumor control

The target cell hypothesis: Munro & Gilbert 1961

- Relevant is the number of tumor stem cells (clonogenic cells) left at the end of treatment
- This is reduced with dose in a manner which accounts for randomness in radiation effects, described by **Poisson statistics**
- The probability of tumor cure depends on the average number of clonogens surviving per tumor

Simulation of a Poisson distribution of surviving cells

0	0	0	2	1	1	1	0	0	0
0	0	0	0	0	0	0	0	1	0
0	0	0	2	1	2	1	2	0	1
1	0	0	0	0	2	1	0	1	2
1	2	0	0	0	0	0	0	0	3
0	0	0	0	0	0	0	1	1	0
1	0	2	0	0	0	0	0	0	1
0	0	0	0	0	3	0	0	1	0
0	3	1	1	1	1	0	0	0	1
0	1	2	1	1	0	0	1	1	0

100 tumors.

Average number of
surviving clonogens per tumor
= 0.5

Each box indicates
the number of surviving clonogens
actually in that tumor

Poisson Statistics – a reminder

In the **Poisson** statistical distribution, the probability $P(x)$ of obtaining x surviving cells per tumor when the mean number of surviving cells per tumor is λ , is:

$$P(x) = \frac{e^{-\lambda} \lambda^x}{x!}$$

Condition: a very, very **large** number of cells in each tumor, but the probability that *any given cell* survives is very, very **small**

Poisson Statistics: local tumor control (“cure”)

Tumor Control Probability, TCP , is the probability of **no** surviving cells in the tumor (*i.e.* $x = 0$).

TCP is therefore given by:

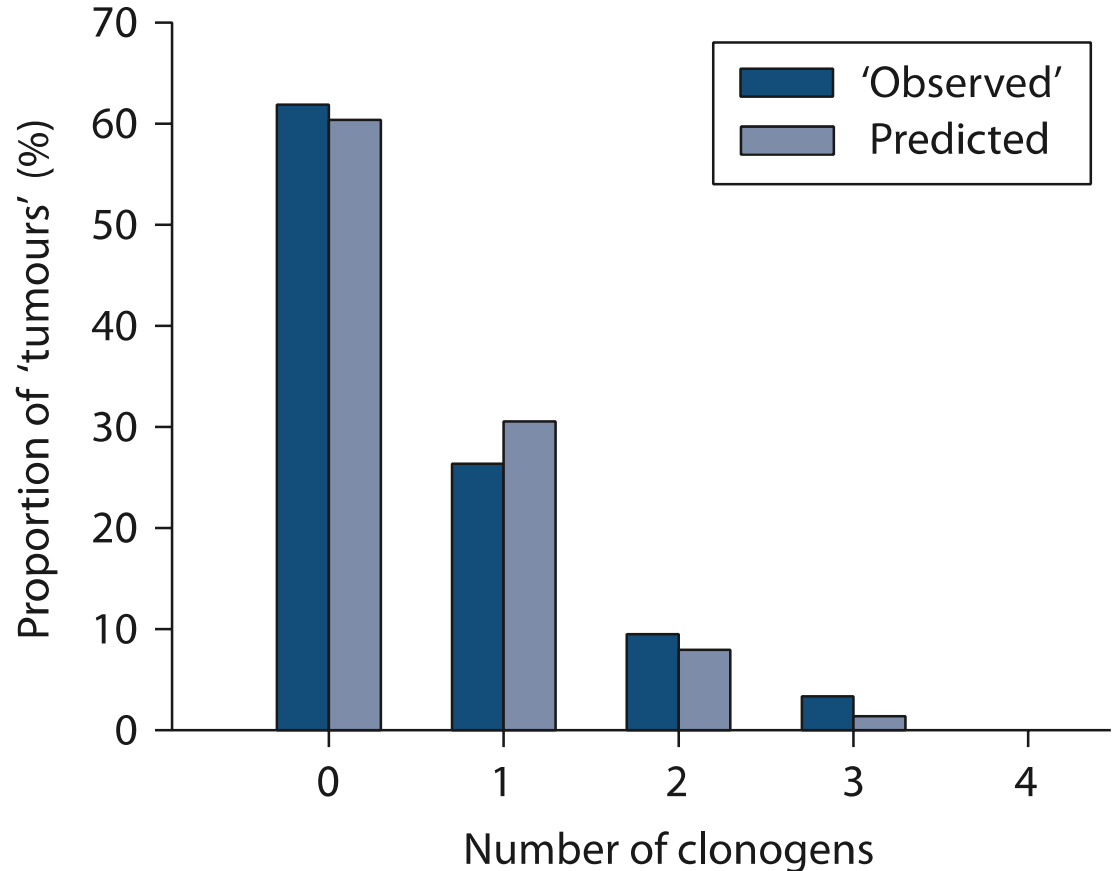
$$TCP = P(0) = \frac{e^{-\lambda} \lambda^0}{0!} = e^{-\lambda} = \exp(-\lambda)$$

λ is the mean number of surviving cells per tumor

Poisson “predicted” versus Monte Carlo “observed”

Average number of
surviving clonogens
= 0.5

Poisson distribution
is confirmed by
“observation”



But λ is a function of dose per fraction, d , and number of fractions, n . Total dose $D = nd$.

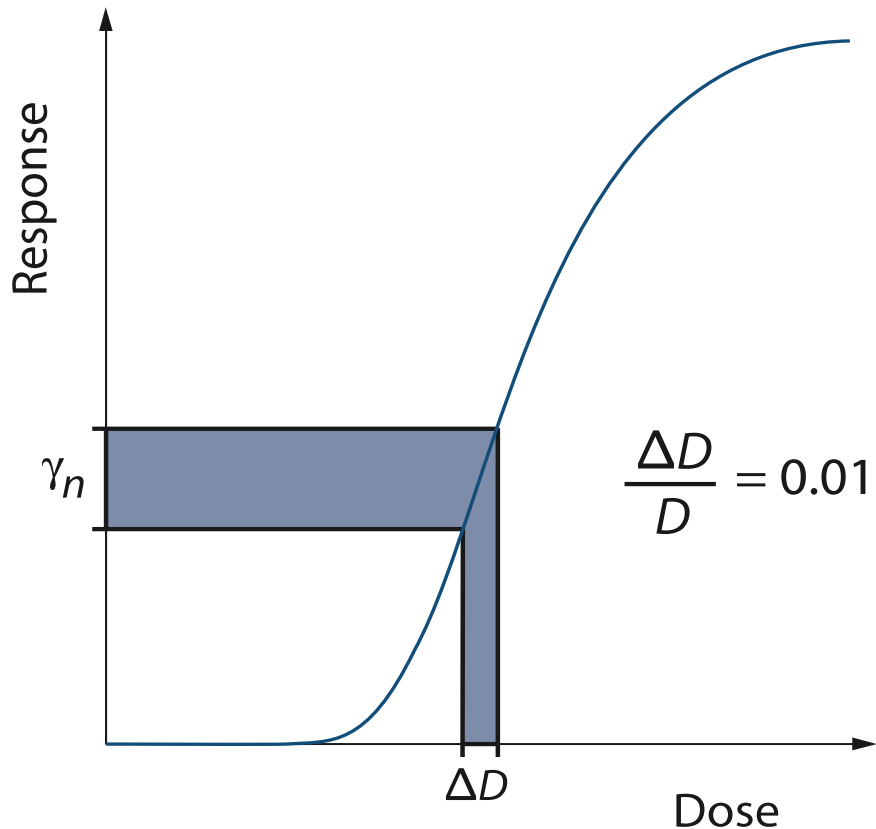
Remember that:

$$S = \lambda/N_0 = e^{-n(\alpha d + \beta d^2)} = \exp(-\alpha D - \beta d D)$$

Therefore:

$$TCP = \exp\left[-N_0 \exp(-\alpha D - \beta d D)\right]$$

Definition of dose-response curve slope



Normalized
dose response gradient, γ :

$$\Delta P \approx \gamma \frac{\Delta D}{D}$$

1% change in dose gives
increase in response = γ %

Usually defined at
the steepest part of curve:
With **Poisson** model,
at Response = **37%** ($0.3679\dots, e^{-1}$)

Interesting consequence of Poisson

It can be shown that:

$$\gamma_{37} = \frac{\ln N_0}{e}$$

This may be used for deducing the number of “tumor clonogens” but any relevance to normal tissue response is doubtful

Logistic model of response

$$P = \frac{\exp(u)}{1 + \exp(u)} \quad u = \ln\left(\frac{P}{1-P}\right)$$

$$u = a_0 + a_1 D + a_2 Dd + \dots$$

$P/(1-P)$ is called the **odds** of the response,
 u is called the **logit** of P

With **Logistic**, the inflection (max slope) occurs at
Response = **50%** ($P = 0.5, u = 0$)

Beware: γ changes with response level

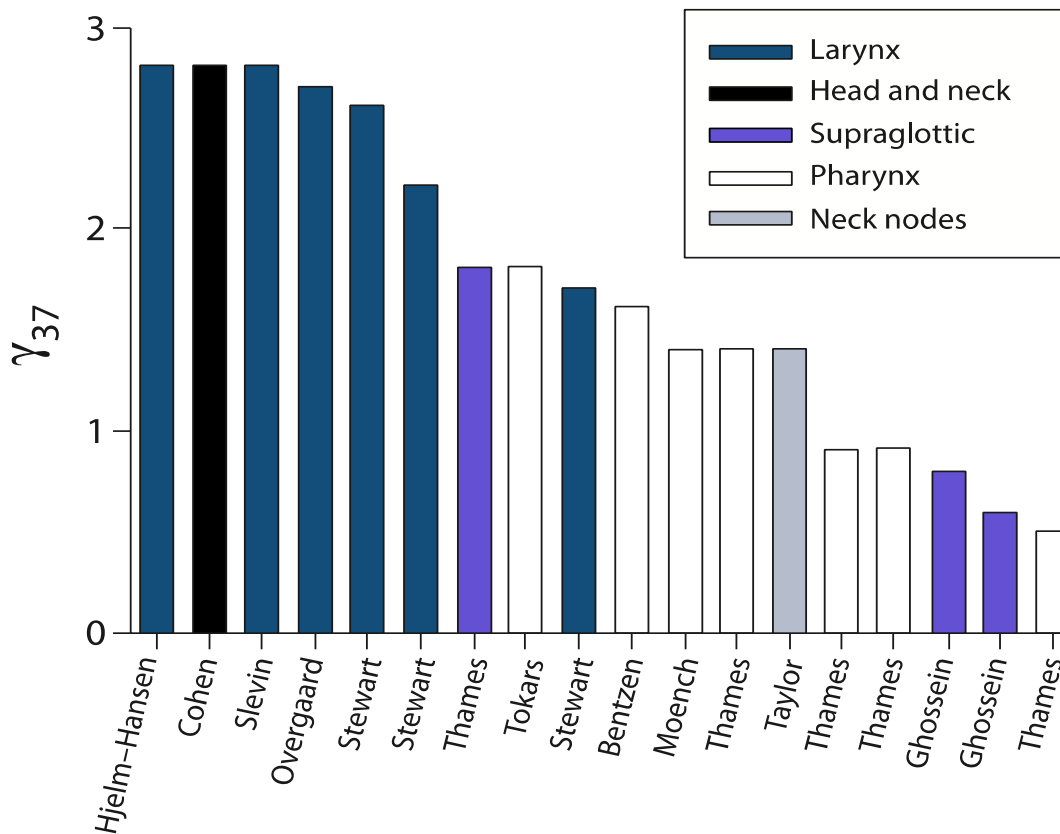
γ_{50}	Response level, %								
	10	20	30	40	50	60	70	80	90
1	0.2	0.4	0.7	0.9	1.0	1.1	1.0	0.9	0.6
2	0.5	1.1	1.5	1.8	2.0	2.0	1.9	1.5	0.9
3	0.9	1.7	2.3	2.8	3.0	3.0	2.7	2.1	1.3
4	1.2	2.3	3.2	3.7	4.0	3.9	3.5	2.8	1.6
5	1.6	3.0	4.0	4.7	5.0	4.9	4.4	3.4	2.0

γ is only useful when you are “on the curve”!

Clinical estimates of γ

Average γ_{37} for
H&N $\approx 2\%$

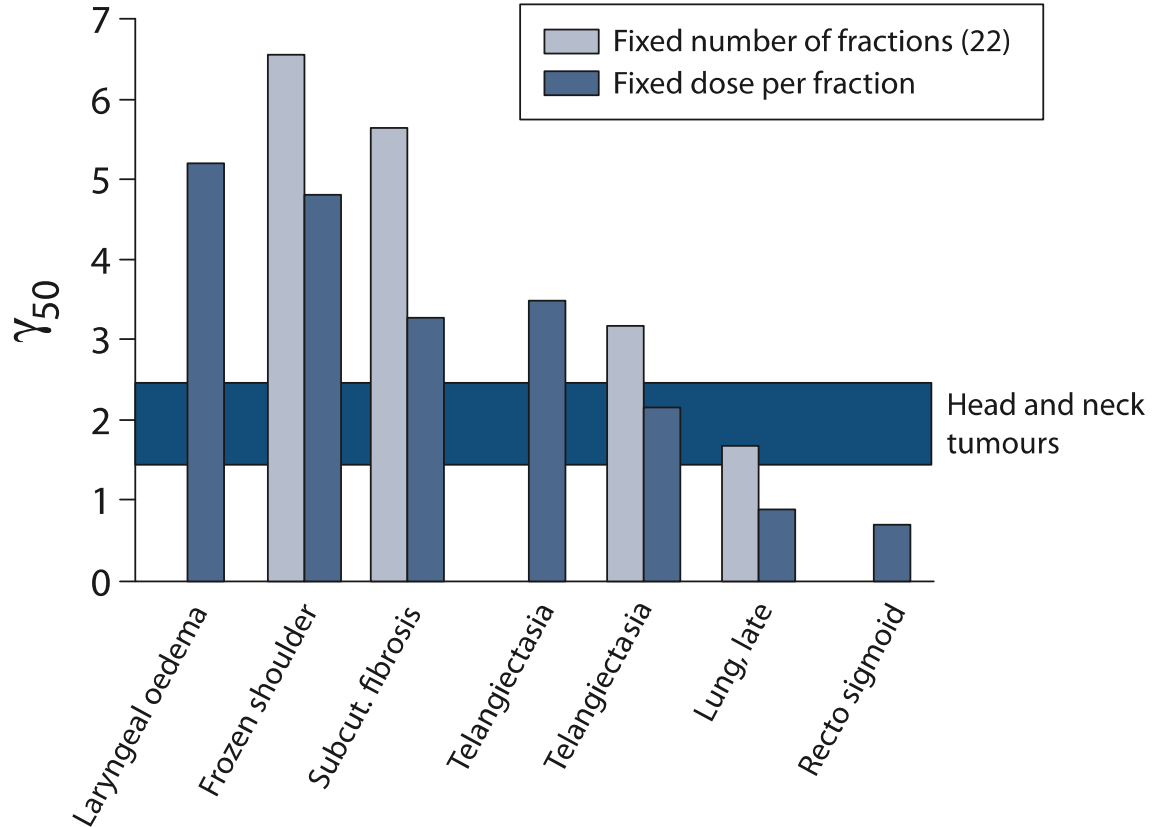
From studies where
dose per fraction
was fixed



Value of γ in some late-reacting tissues

Compared with tumors, γ is usually larger

Dose response curves can be steeper, more so when fixed fraction number, *i.e.* higher dose per fraction

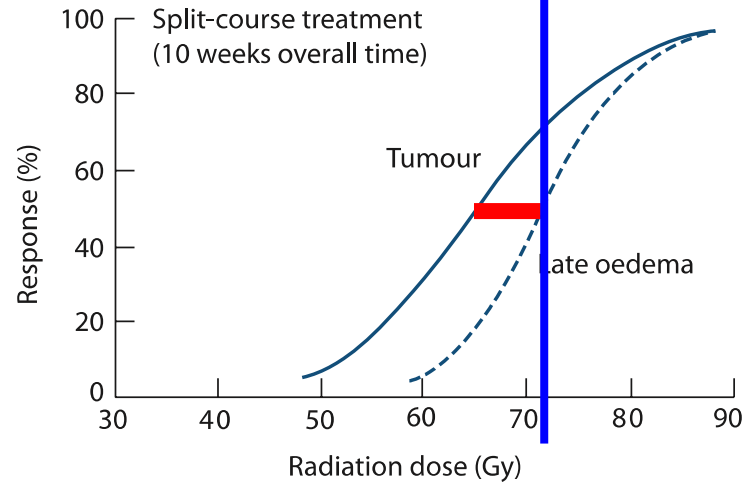
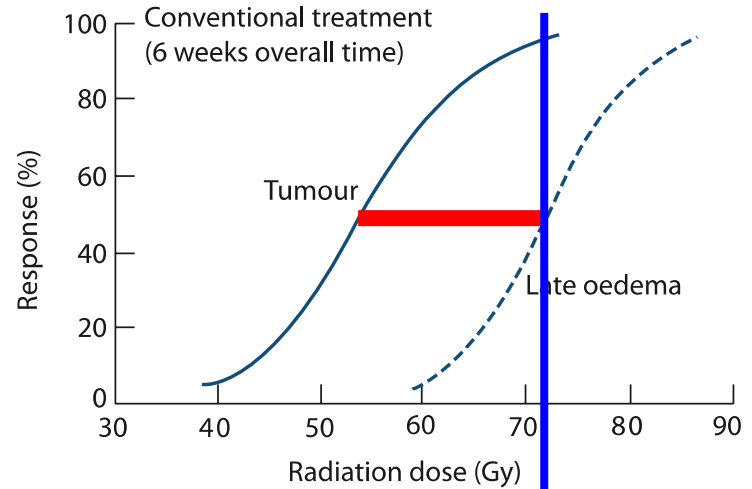


Bentzen (1994)

Bentzen and Overgaard (1996)

Balancing risks and benefits: The therapeutic window

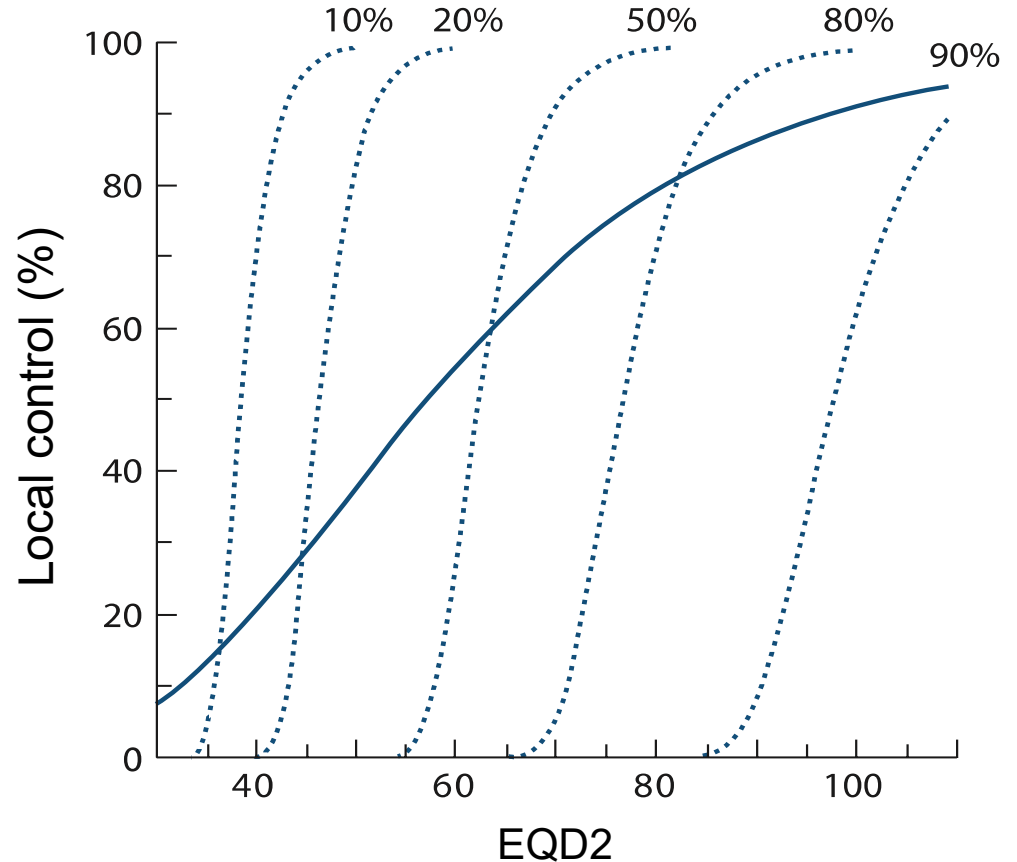
Example: protraction of overall treatment time is detrimental!



Modifying the steepness of the dose-response

Oropharyngeal
cancer

Homogeneous
patient populations with
radiosensitivity equal to
selected percentiles of
radiosensitivity
distribution in total
population



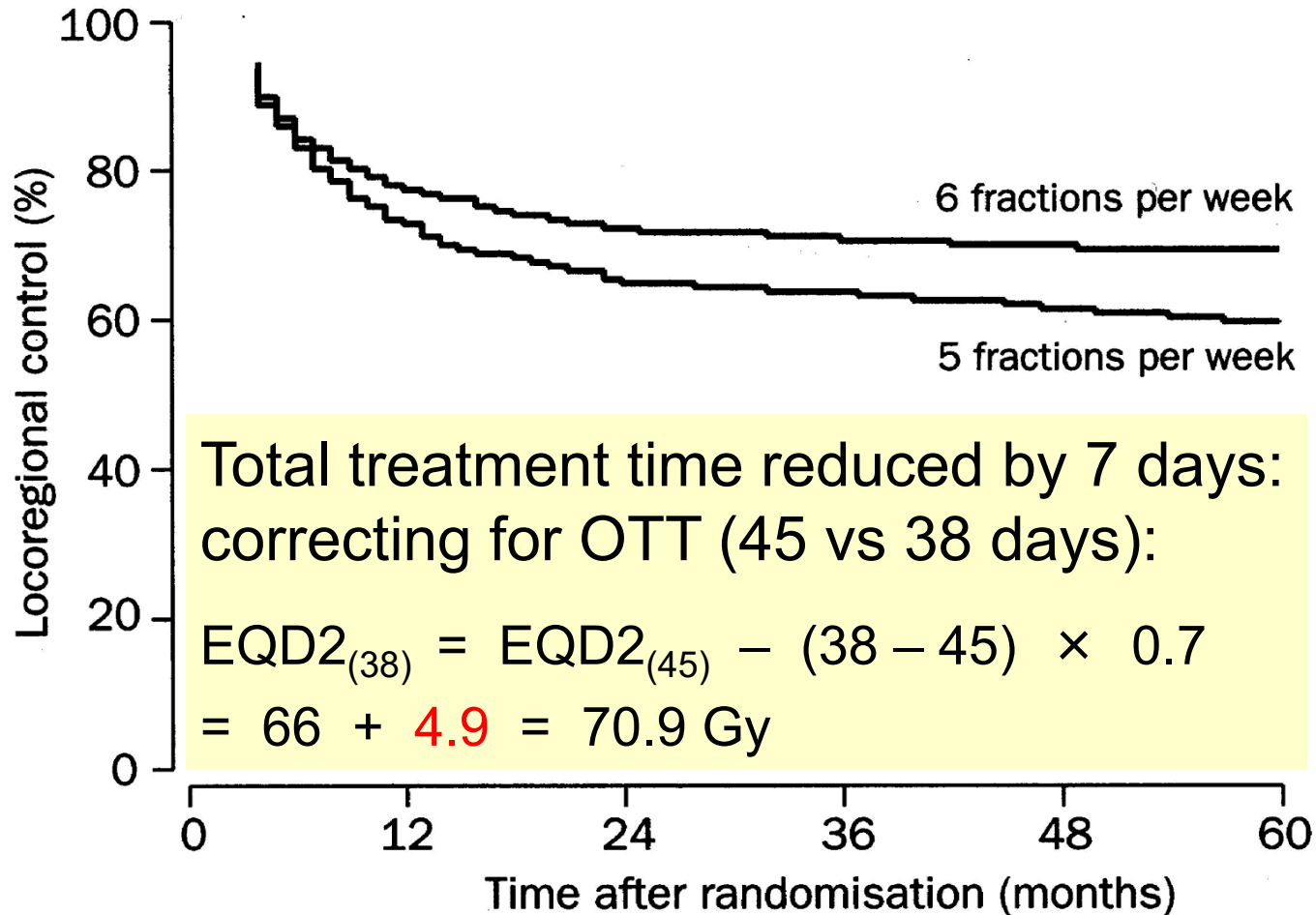
Clinical data to test modeling

👁 Five compared with six fractions per week of conventional radiotherapy of squamous-cell carcinoma of head and neck: DAHANCA 6&7 randomised controlled trial

Jens Overgaard, Hanne Sand Hansen, Lena Specht, Marie Overgaard, Cai Grau, Elo Andersen, Jens Bentzen, Lars Bastholt, Olfred Hansen, Jørgen Johansen, Lisbeth Andersen, Jan F Evensen, on behalf of the Danish Head and Neck Cancer Study Group

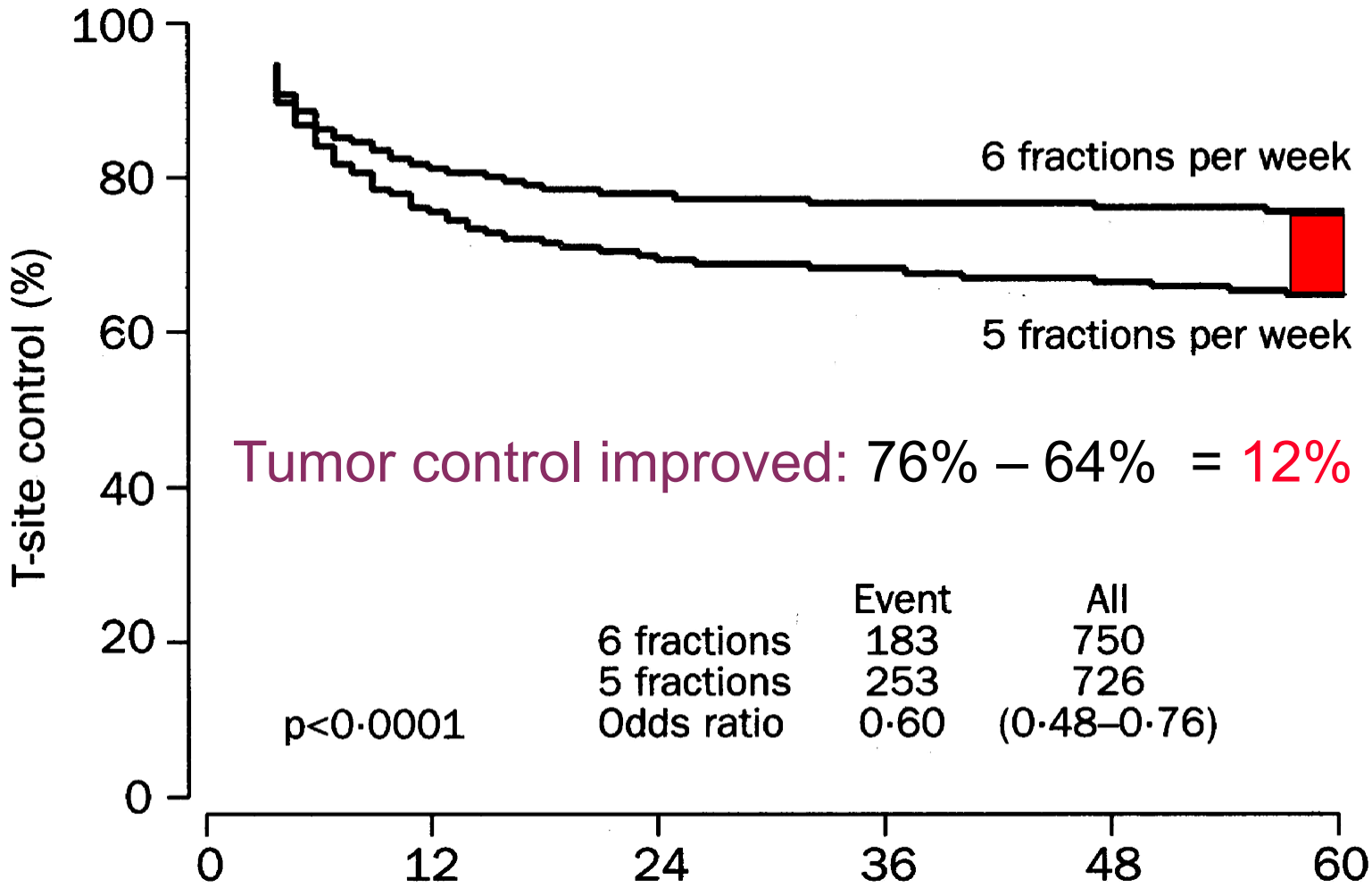
Lancet 2003;362:933-40

Convert from a change in **dose**
to a change in **response rate**



From change in dose to change in RR

$$\begin{aligned}\Delta R &\approx \gamma \times \frac{\Delta D}{D} \times 100\% \\ &= 1.6 \times \frac{4.9}{66} \times 100 = 12\%\end{aligned}$$



Tumor control improved: 76% - 64% = 12%

Clinical manifestations of normal tissue damage

Remember...

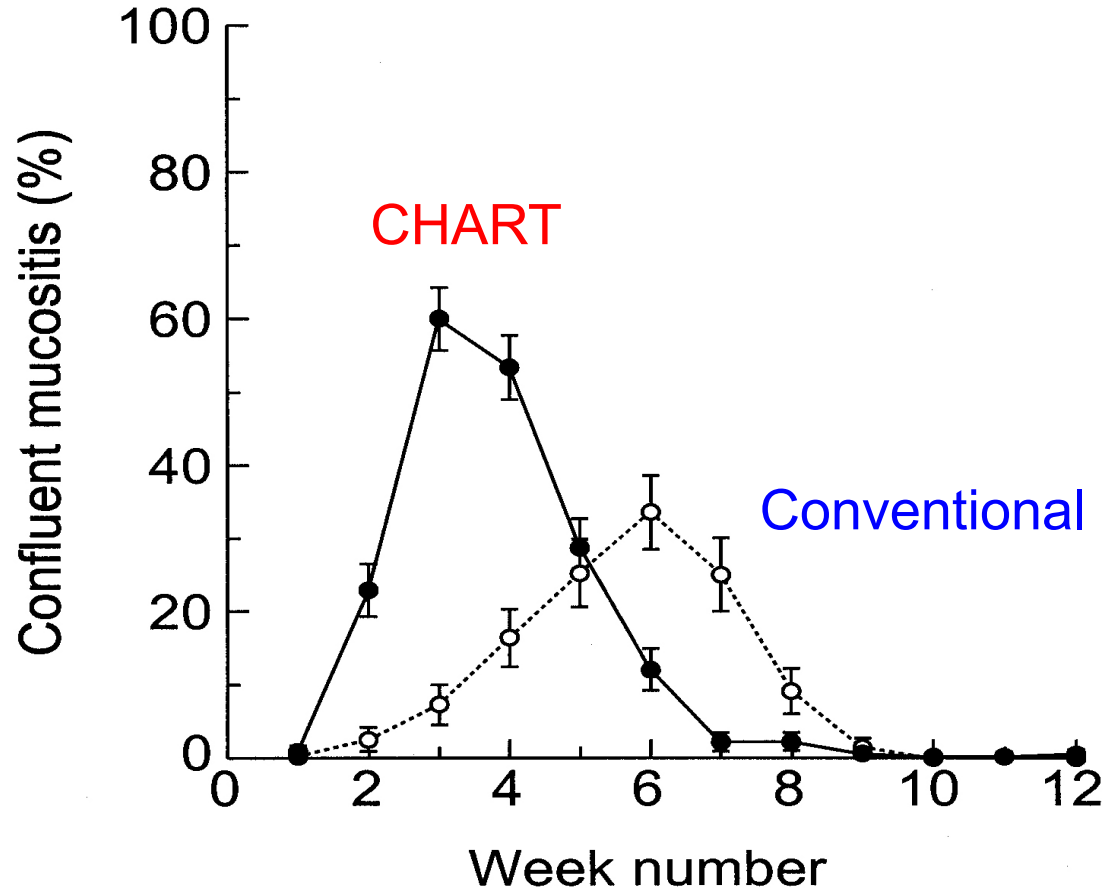
Document *tumor* response

Document *normal tissue* reactions

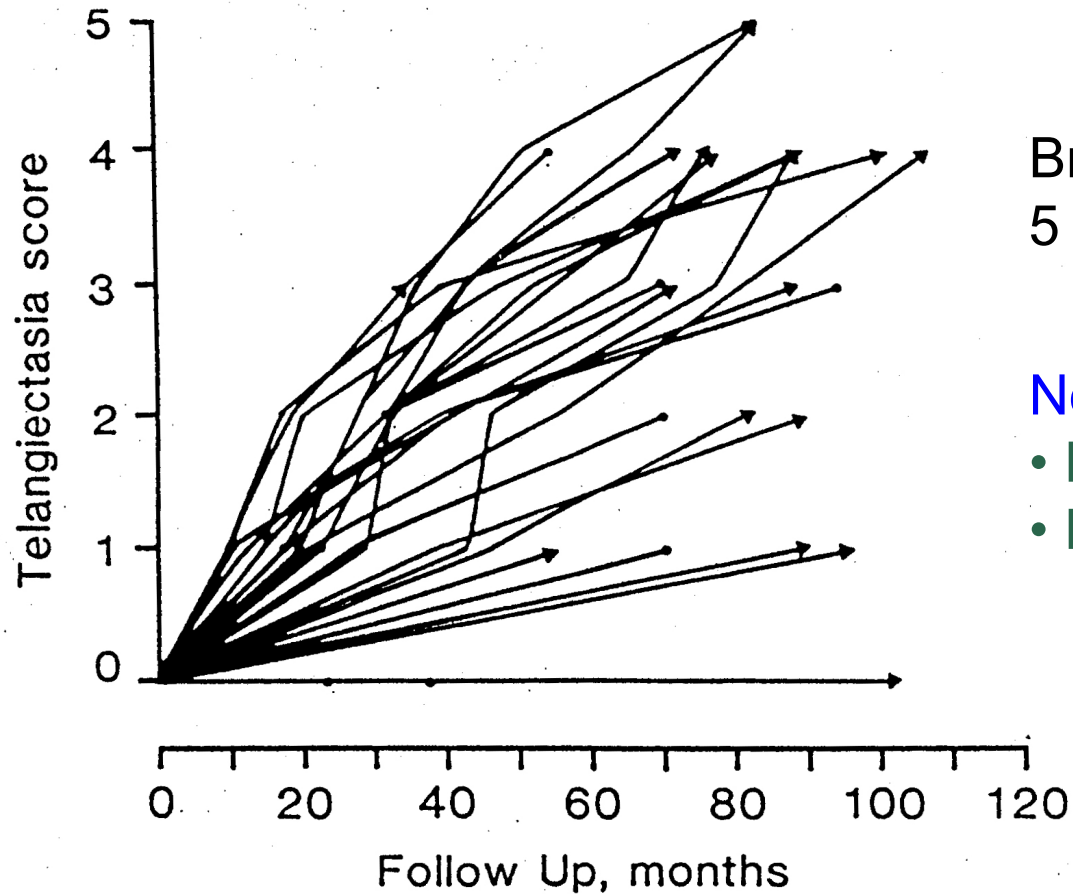
Three types of clinical toxicity data

Type of end-point	Statistical data type	Scoring system	Examples
Binary	Categorical all-or-nothing response	Yes / No	Radiation-induced second tumors
Graded	Ordinal ranking of severity	<i>e.g.</i> None / mild / moderate / severe	Telangiectasia; subcutaneous fibrosis
Continuous	Continuous	“Laboratory value”	Kidney ⁵¹ Cr-EDTA clearance; CT density of pulmonary fibrosis

Prevalence of confluent mucositis



Progressive nature of late reactions

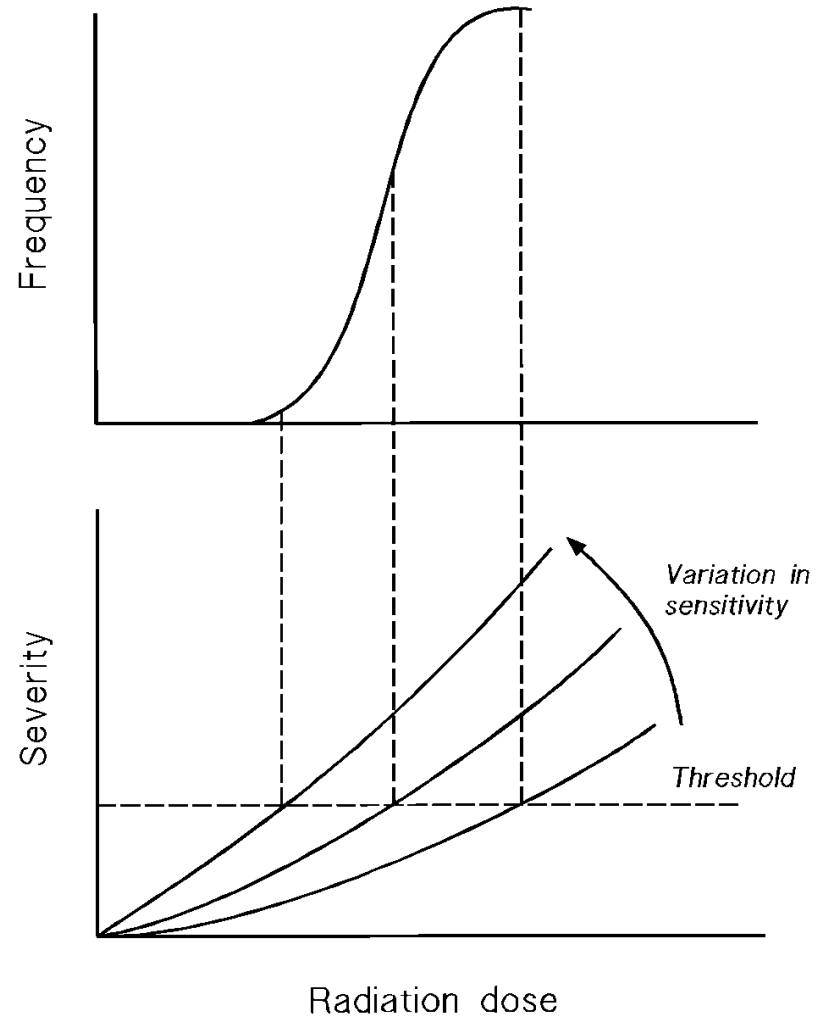


Breast Ca, post-op R/T
5 × 1.8 Gy/w, N = 35

Note:

- Long latent times
- Large inter-individual variation

Conversion of continuous
(deterministic or non-stochastic)
into all-or-nothing
(stochastic)
responses



Crude versus actuarial estimates of complications

End-point	Primary tumour	Crude estimate	Actuarial	Remarks	Reference
Radiation myelopathy	Lung	4 ± 1% at 3 years	30 ± 15%	Median survival 9 months	Hatlevoll
Marked telangiectasia	Breast	39 ± 6%	62%	Follow-up 1.5-6 years; long latent period	Bentzen
Severe rectosigmoid complications	Uterine cervix: FIGO IIb FIGO IIIb+IVa	9 ± 5% 15 ± 3%	10 ± 6% at 5y 39 ± 8% at 5y	Patients with IIIb and IVa disease received a higher dose but had fewer complications	Unpublished

Conclusion: Always quote actuarial values

https://ctep.cancer.gov/protocolDevelopment/electronic_applications/docs/

Common Terminology Criteria for Adverse Events (CTCAE)

Version 5.0

Published: November 27, 2017

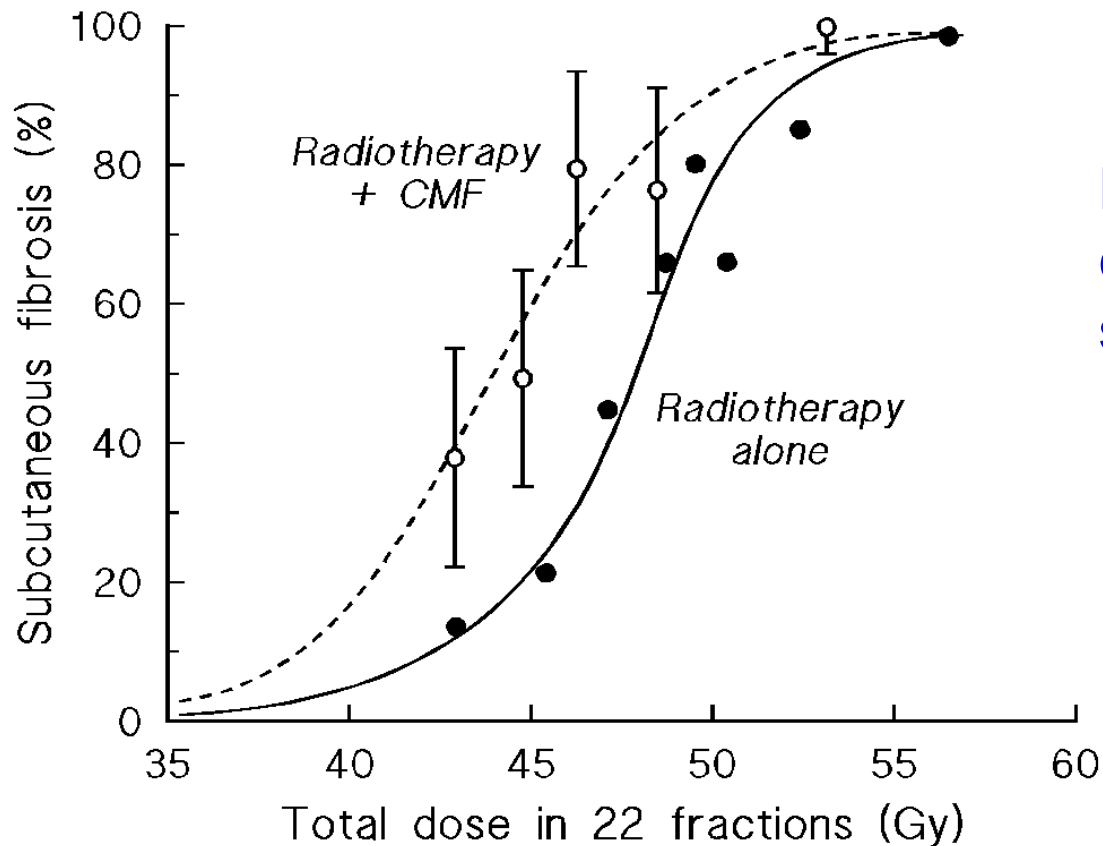
[CTCAE_v5_Quick_Reference_8.5x11.pdf](#)

U.S. DEPARTMENT OF HEALTH AND HUMAN SERVICES

National Institutes of Health

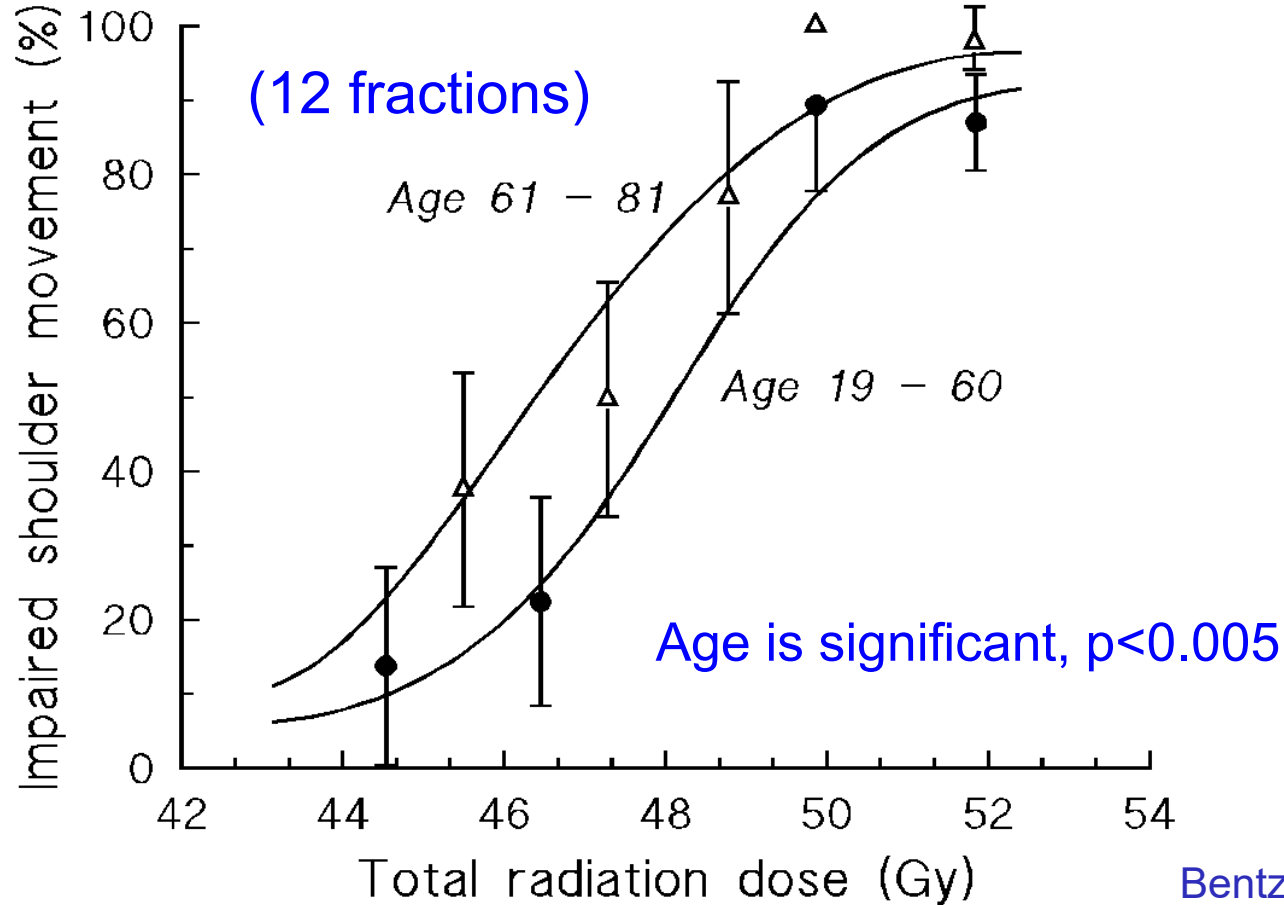
National Cancer Institute

Treatment-related factors: influence of adjuvant chemo in post-mastectomy R/T



No such change with
cyclophosphamide as
single agent

Impaired shoulder movement after post-mastectomy R/T



Dose-volume models for normal tissues

- Predicting normal tissue toxicity has become more complicated by the use of IMRT, non-uniform dose distributions and partial organ irradiation
- Mathematical and biophysical models are developed to describe late normal tissue toxicity
- Toxicity is assessed from the complete dose distribution throughout an OAR in an integrative manner

NTCP models

Example:

The Lyman model of dose-volume effects in normal tissue

- Relates NTCP to dose and volume irradiated
- Assumes a normal distribution of complications as a function of dose for each uniformly irradiated fractional organ volume

Lyman model of dose-volume effects in normal tissue

$$NTCP(D, v) = \frac{1}{\sqrt{2\pi}} \cdot \int_{-\infty}^{u(D, v)} \exp(-x^2/2) dx$$

$$u(D, v) = \frac{D - D_{50}(v)}{m \cdot D_{50}(v)}$$

$$0 < n < 1$$

Larger n , more volume effect

$$D_{50}(v) = \frac{D_{50}(1)}{v^n}$$

(see *BCR5 book, Ch 7.6*)

D_{50} = uniform dose producing 50% incidence of specific effect

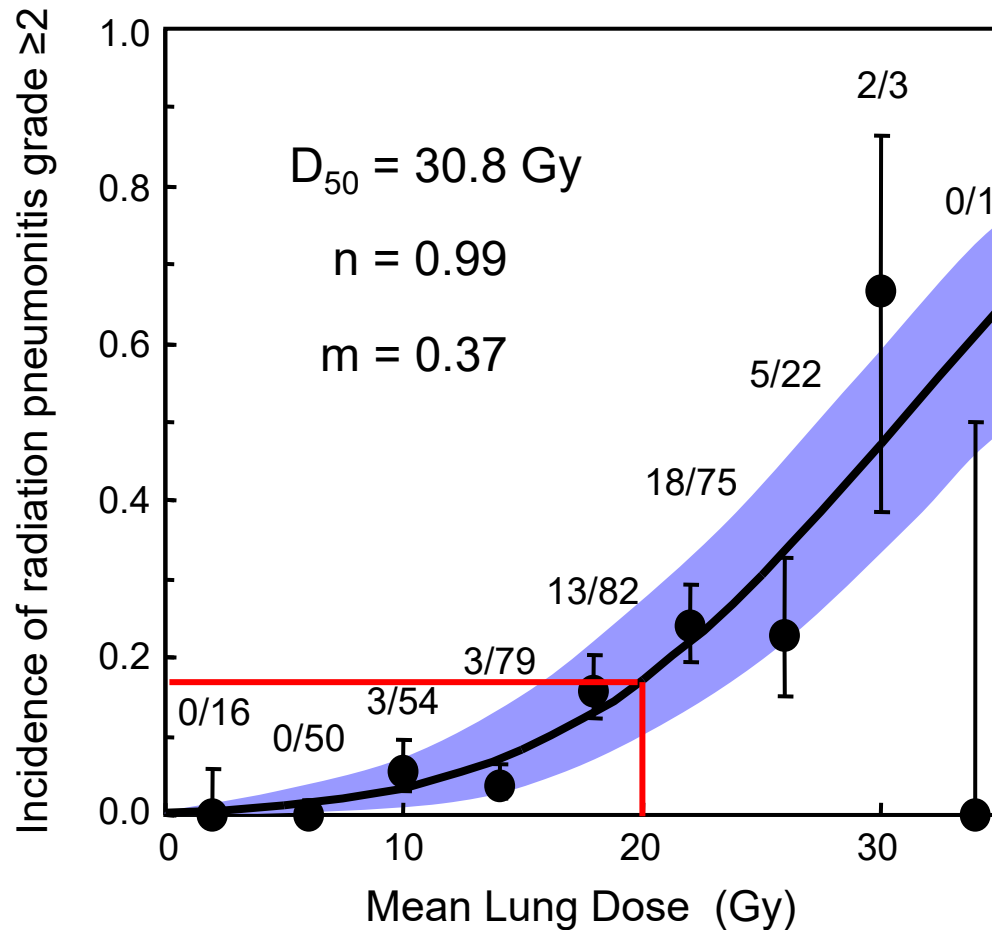
n = denotes influence of volume effect in organ of interest

m = inverse of dose response curve gradient

NTCP models

Organ	Toxicity	D₅₀	Volume effect (n)	Dosimetric descriptor
Parotid gland	Xerostomia	28.4 Gy	large (1)	mean dose
Lung	gr ≥ 2 pneumonitis	30.8 Gy	large (0.99)	V20, MLD
Heart	RIHD		intermediate (0.35–0.64)	Vd, MHD
Spinal cord	myelopathy		marginal (except very small volumes)	EQD2
Liver	RILD	40-45 Gy	large (0.69–0.97)	MLD, Vd
Rectum	proctitis, ulceration	80 Gy	small (serial)	V70, V50

Complications versus mean lung dose



Seppenwoolde et al.
Int J Radiat Oncol Biol Phys 2003;55:724-35

Summary

- Dose-response data usually defined in terms of probability
- Steepness of the dose response at a defined level can be used to convert change in dose to response
- Dose-response curves for normal tissues are steeper than those for tumors
- Heterogeneity in population data (particularly tumors) tend to make dose-response curves less steep
- NTCP models are not well validated and **require caution** when applied to clinical data; simpler dosimetric descriptors may be more useful