

Statistical Challenges in Immuno-Oncology

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Outline

Cytotoxic vs. cytostatic agents

- ◆ Mechanism of action
- ◆ Endpoints

Immunotherapies

- ◆ Important issues to consider in study design and analysis
- ◆ Efficacy
 - Overall Survival

MOA: Cytotoxic vs. Cytostatic Agents

Cytotoxic agents

- ◆ Dose-dependent rapid cell kill or tumor shrinkage
- ◆ Lack of selectivity leads to undesired toxicity or side effects

Cytostatic agents

- ◆ Inhibit or suppress cellular growth or division which leads to delayed progression
- ◆ Minimal or less severe toxicity, prolonged duration of treatment at lower dose

Endpoints: Cytotoxic vs. Cytostatic

Cytotoxic

- ◆ OS: Clinical benefit
- ◆ BOR (WHO or RECIST): Direct cell kill action leads to tumor shrinkage

Cytostatic

- ◆ OS: Clinical benefit
- ◆ PFS/TTP: Stop or delay tumor growth
- ◆ BOR: Some may shrink tumor

Immunotherapies

Stimulate the patient's own immune system to fight cancer

- ◆ Immune cell activation; change in tumor burden
- ◆ Toxicity or side effects caused by the modulation of immune activity

Endpoints remain similar

- ◆ OS: clinical benefit
- ◆ BOR: tumor shrinkage

Important Issues in Design and Analysis in Immuno-Oncology

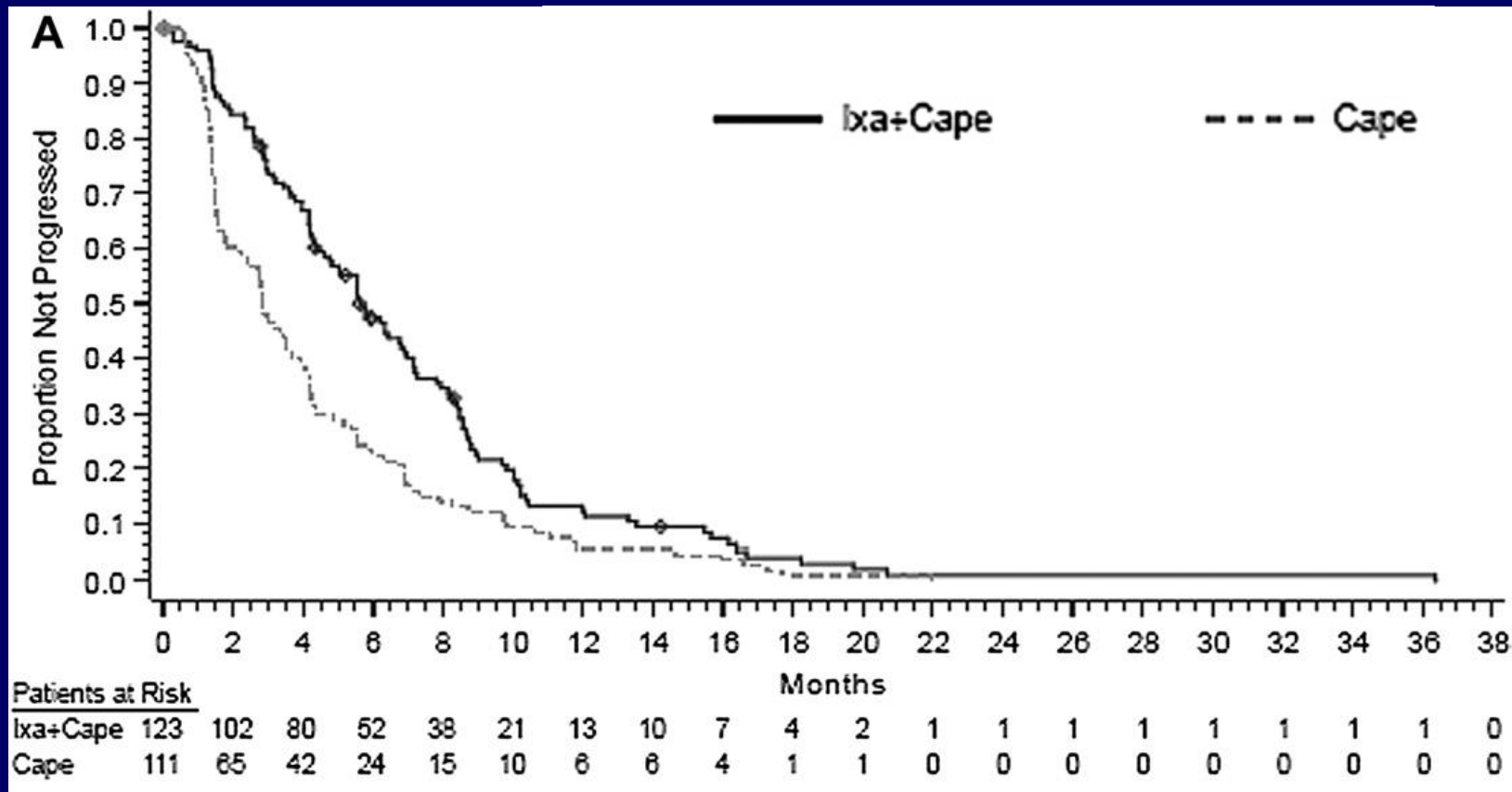
Sample size determination

- ◆ Expected number of events
- ◆ Timing of analysis

Efficacy analysis

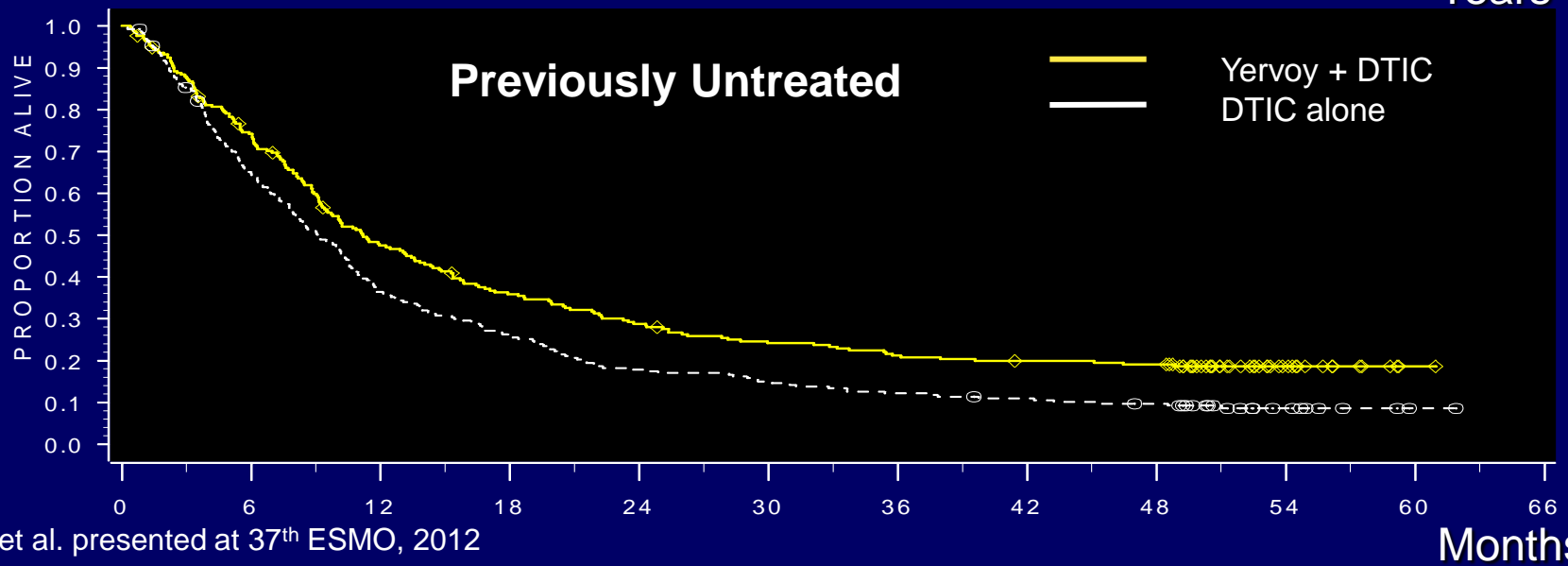
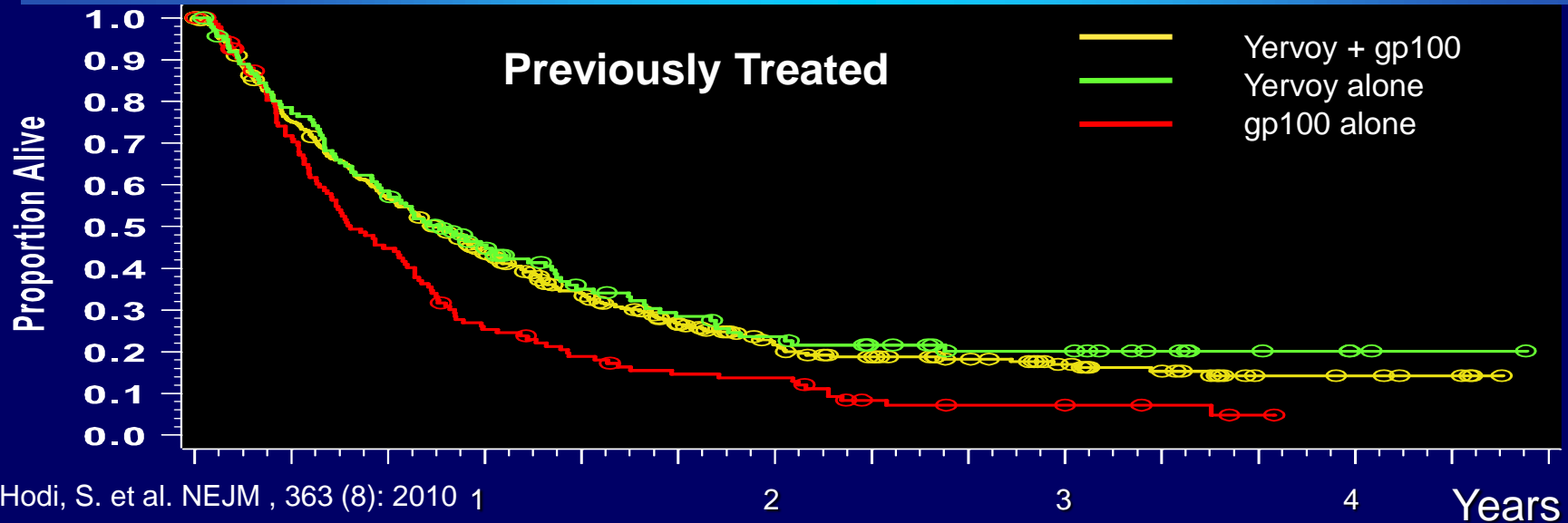
- ◆ Interim analysis strategy
- ◆ Additional analysis considerations

Typical Survival Curve – Advanced Breast Cancer

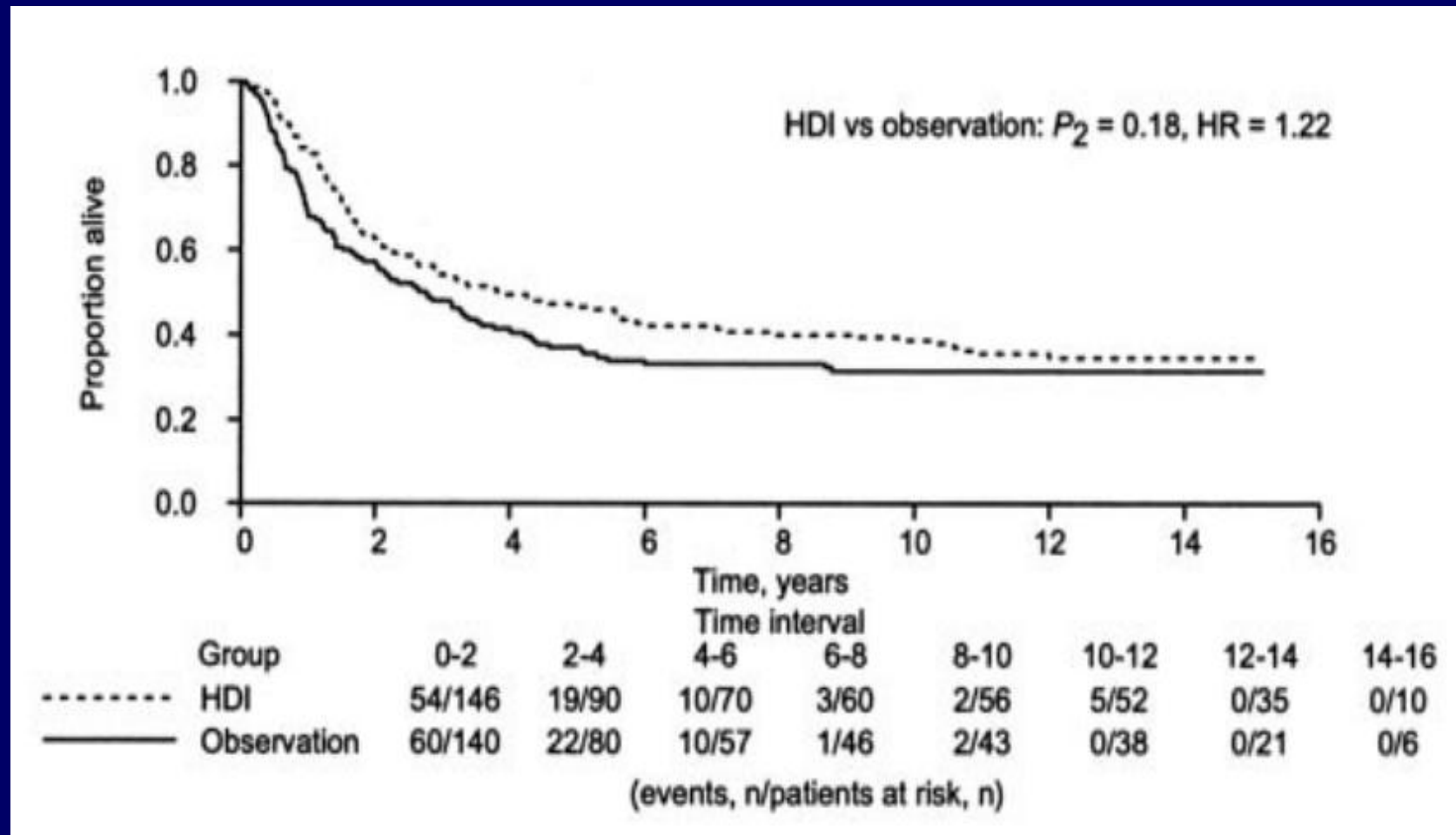


J. Jassem et al. / The Breast 21 (2012) 89–94

Ipilimumab (Yervoy) in Metastatic Melanoma

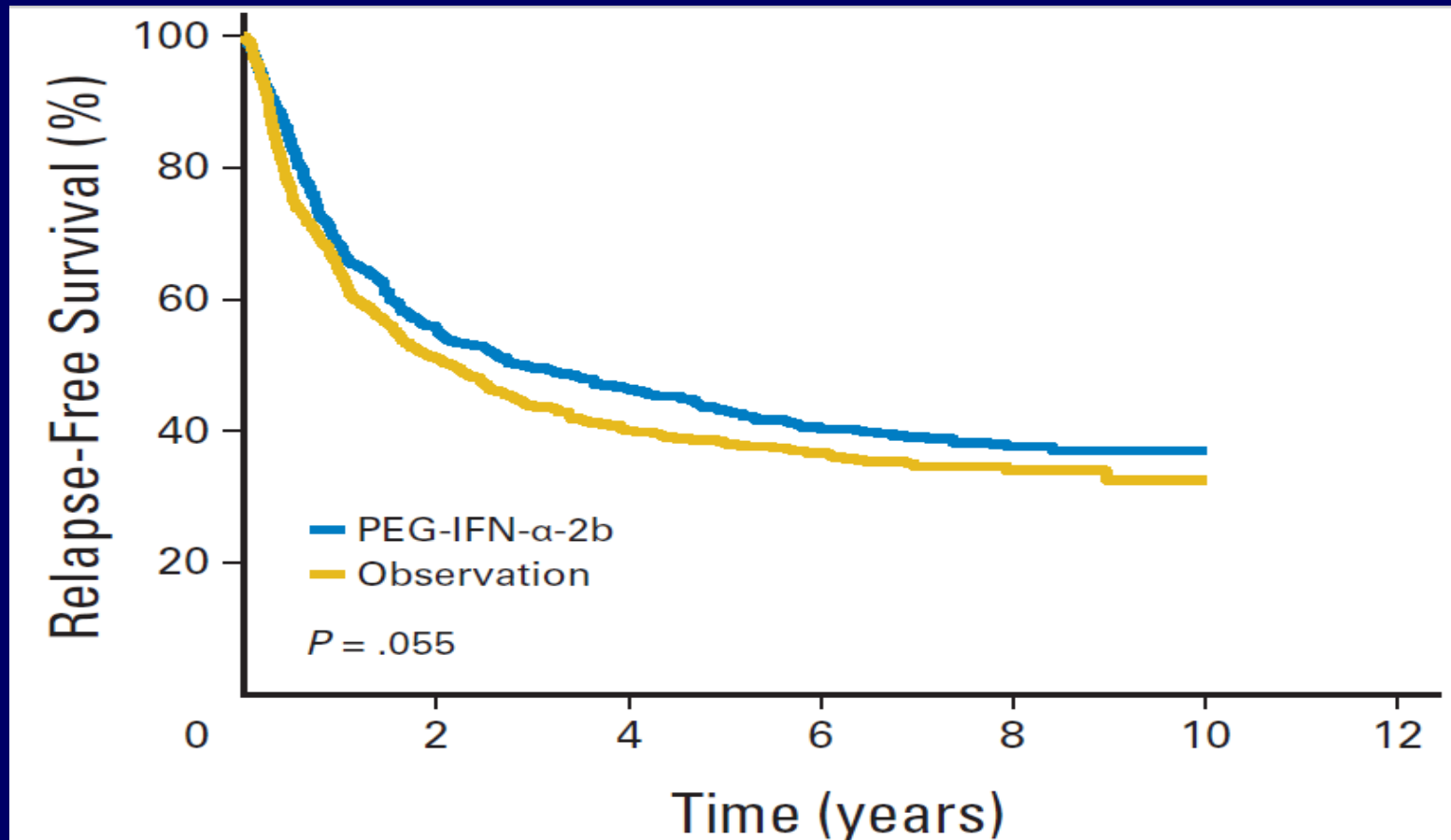


Interferon alfa-2b (Intron-A) – Adjuvant Melanoma



Kirkwood et al., 2004, *Clinical Cancer Research*

Pegylated Interferon alfa-2b (Sylatron): Relapse-Free Survival – Adjuvant Melanoma



Eggermont, AMM, et al., 2012, *Journal of Clinical Oncology*

Study Design and Sample Size Determination

Standard study design

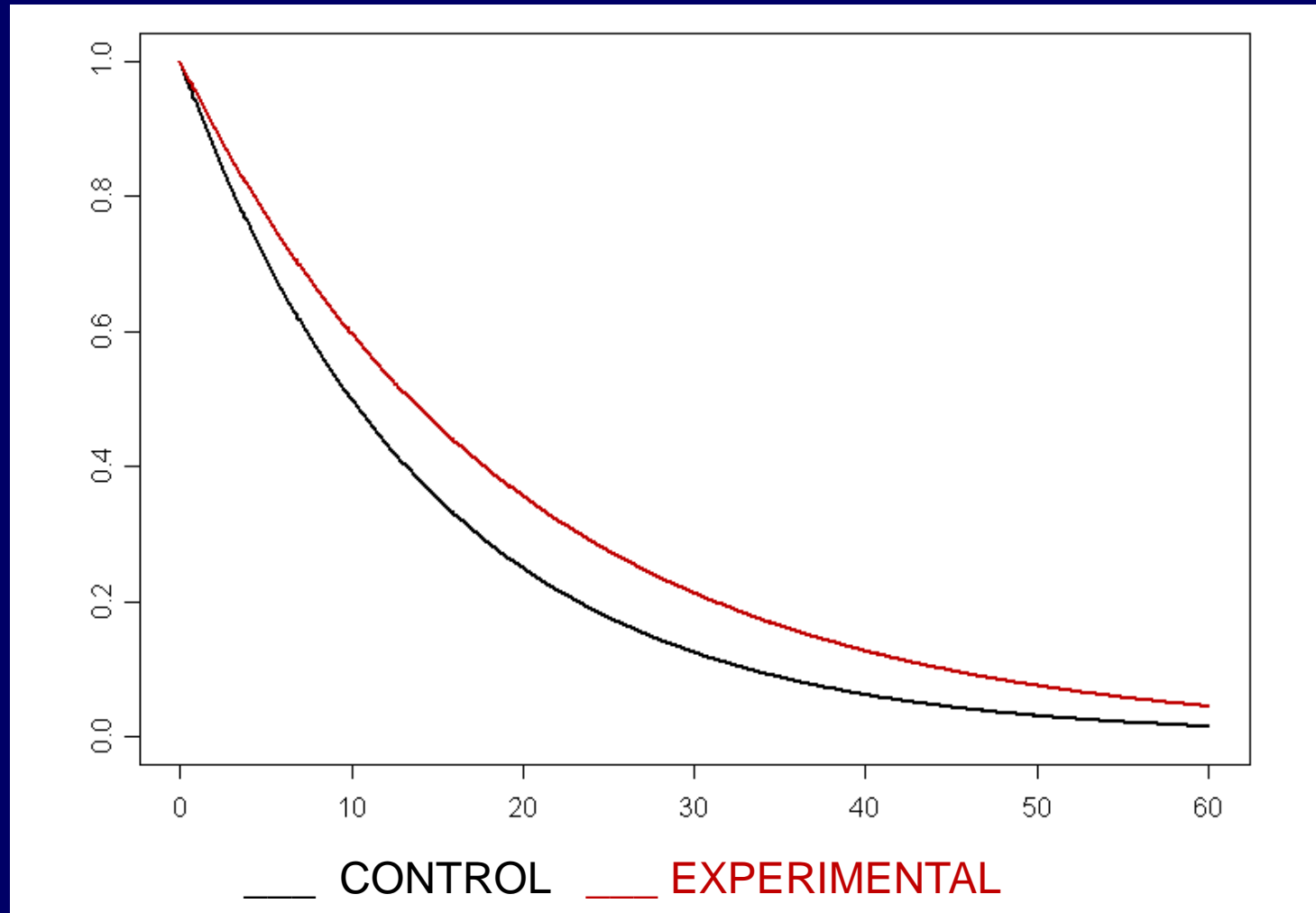
- ◆ Assumes exponential distribution

Unconventional study design

- ◆ Long-term survival (or “cure rate” or “functional cure”)
- ◆ Delayed clinical effect

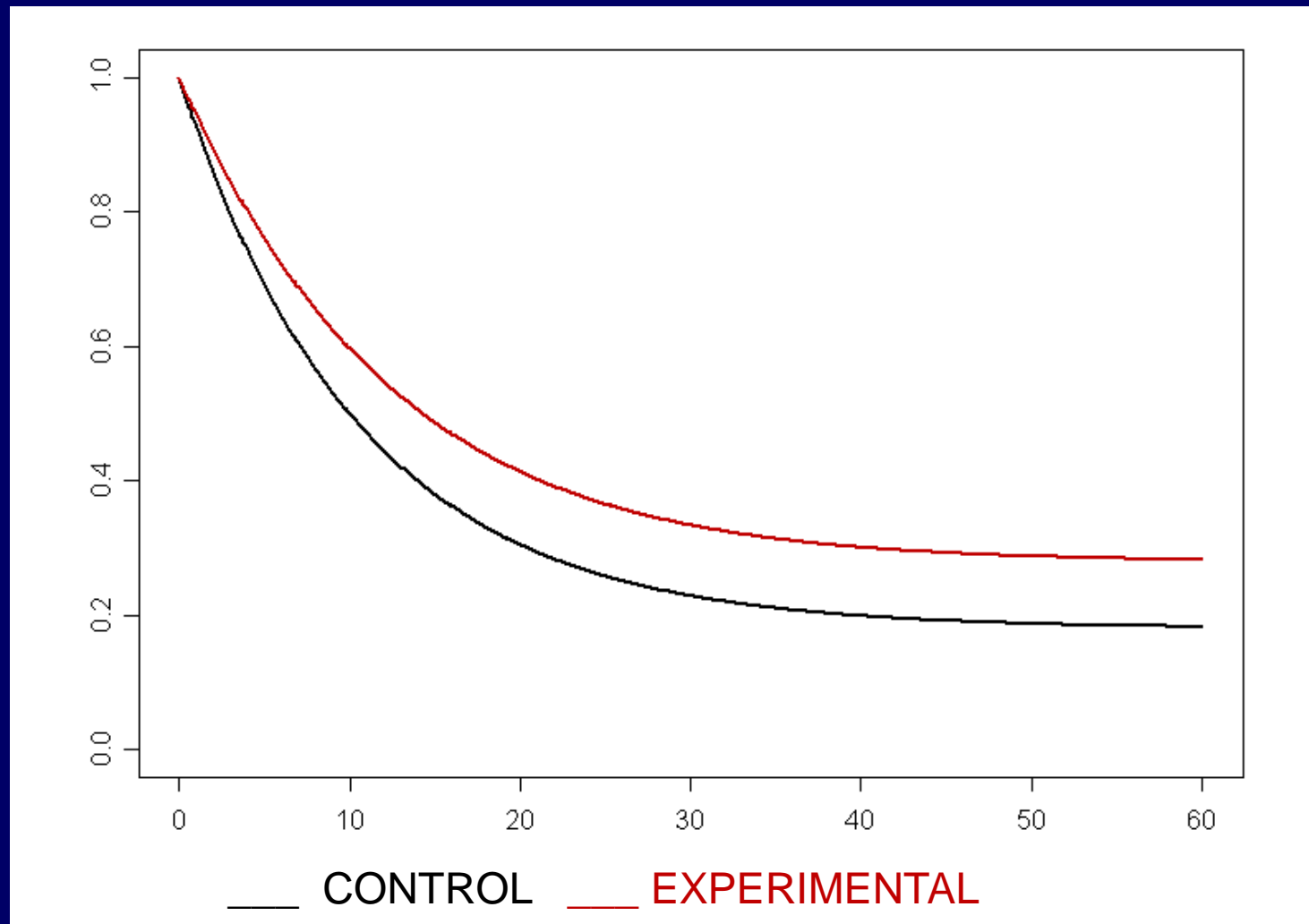
Does unconventional study design impact sample size / power calculation?

Exponential OS Study Design



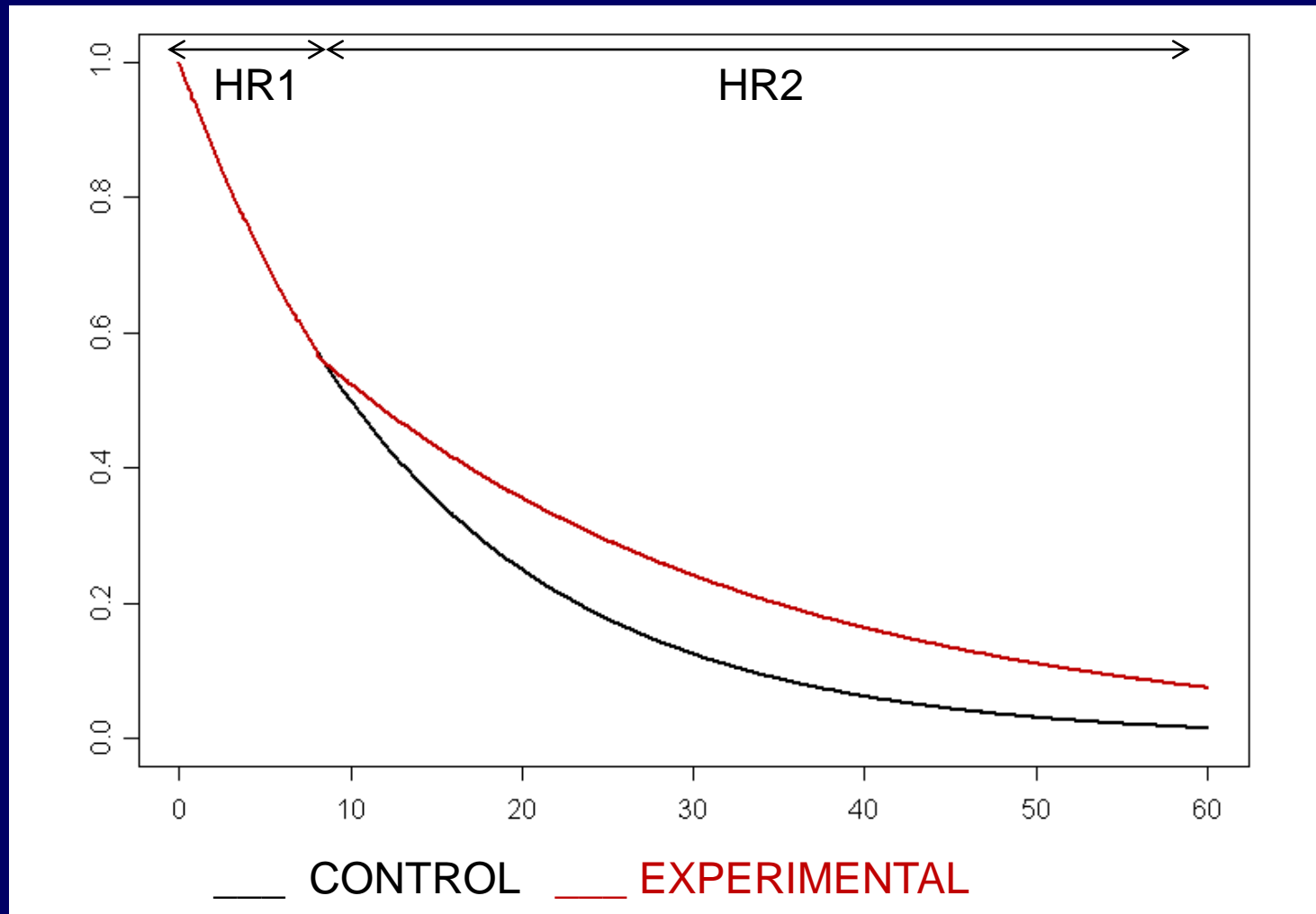
* Proportional hazards model (exponential)

Long-Term (LT) Survival



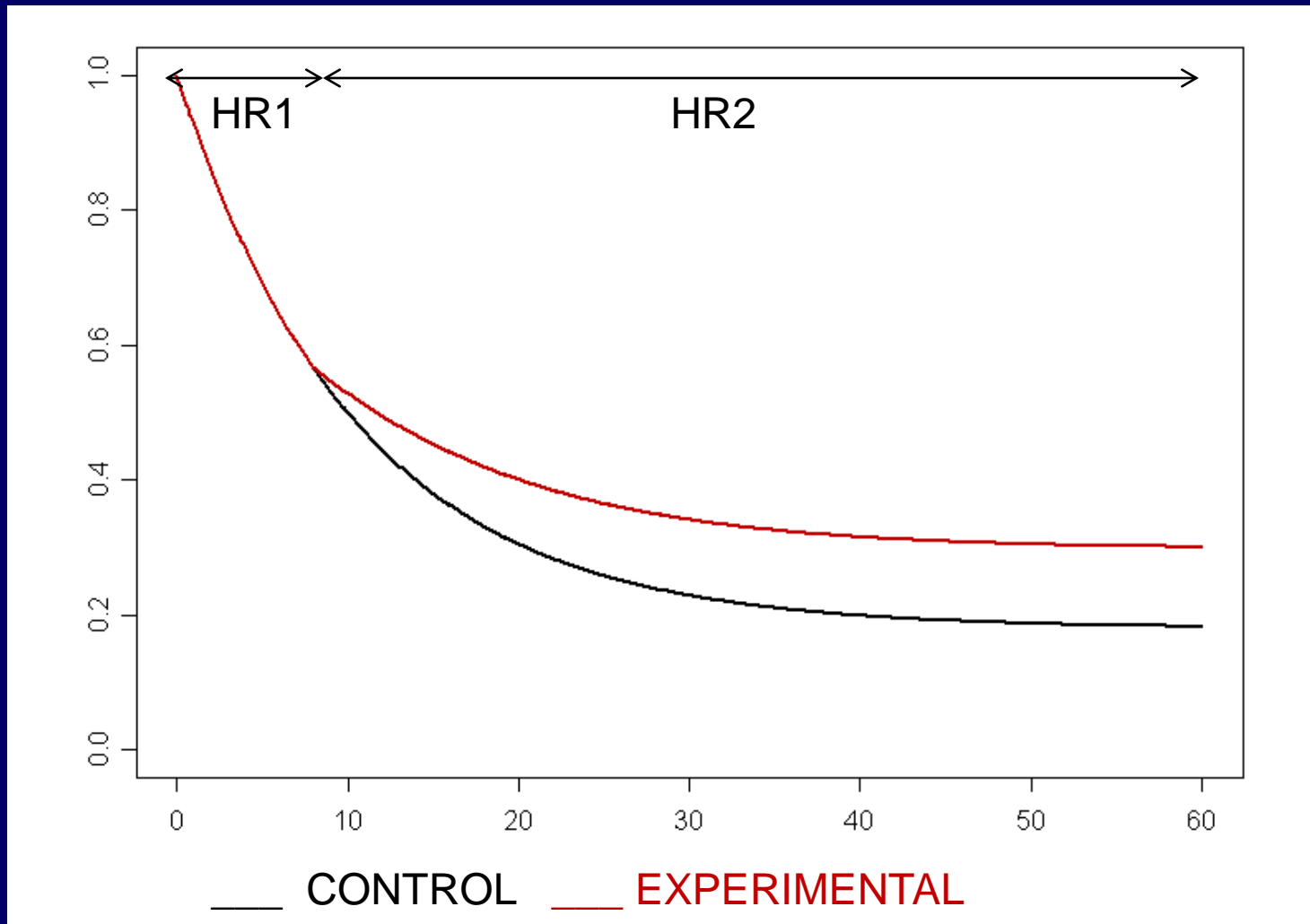
* Proportional hazards cure model

Delayed Clinical Effect



* Non-proportional hazards model

Delayed Clinical Effect with Long-Term Survival



* Non-proportional hazards cure model

Example of a Standard Study Design

Consider the following standard study design

- ◆ Exponential distribution
- ◆ Median OS: 12 vs. 16 months (HR=0.75)
- ◆ Power: 90%
- ◆ Two-sided type I error rate: 5%
- ◆ Accrual rate: 20 pts/month
- ◆ No interim analysis
- ◆ Required number of events: 512 events
- ◆ Sample size: 680 subjects
- ◆ Accrual duration: 34 months
- ◆ Study duration: 48 months

Impact of LT Survival and Delayed Clinical Effect on Study Duration and Power

	Standard (exponential)	LT Survival	Delay	LT Survival / Delay
LT survival	--	0.10/0.18	--	0.10/0.17
Delayed effect	--	--	3 m	3 m
Sample size	680	680	680	680
# events	512	512	512	512
Hazard Ratio	0.75	0.75	1/0.75	1/0.75
Power	0.90	0.90	0.70	0.70
Study duration	48	55	47	54

* Based on 10000 simulations

Impact of LT Survival and Delayed Clinical Effect on Study Duration and Power

Long-term survival

- ◆ Results in prolonged study duration
- ◆ Higher LT survival results in longer study duration

Delayed clinical effect

- ◆ Reduces statistical power
- ◆ Longer delay results in more power loss

Expected number of events

- ◆ Can the number of events be achieved?

Follow-up duration

- ◆ Is the study designed to allow sufficient follow-up for all patients?

Interim Analysis Strategy

Necessity of interim analysis

- ◆ Interim analysis vs. final analysis only

Timing of interim analyses

- ◆ Early vs. late interim analysis

Type of interim analysis

- ◆ Superiority vs. futility

Probabilities for Stopping at Interim Analysis

	Standard (exponential)	LT Survival	Delay	LT Survival / Delay
Interim sample size	520	540	480	500
# events	256	256	256	256
PET _a (superiority)	0.25	0.25	0.06	0.06
PET _a (futility)	0.01	0.01	0.08	0.08

PET_a = Probability of Early Termination when agent is active

Using O'Brien-Fleming boundaries

* Based on 10000 simulations

Interim Analysis Strategy - Conclusion

Delayed clinical effect and LT survival

- ◆ Careful consideration warranted:
 - Necessity of interim analysis
 - Timing of interim analysis
 - Type of interim analysis

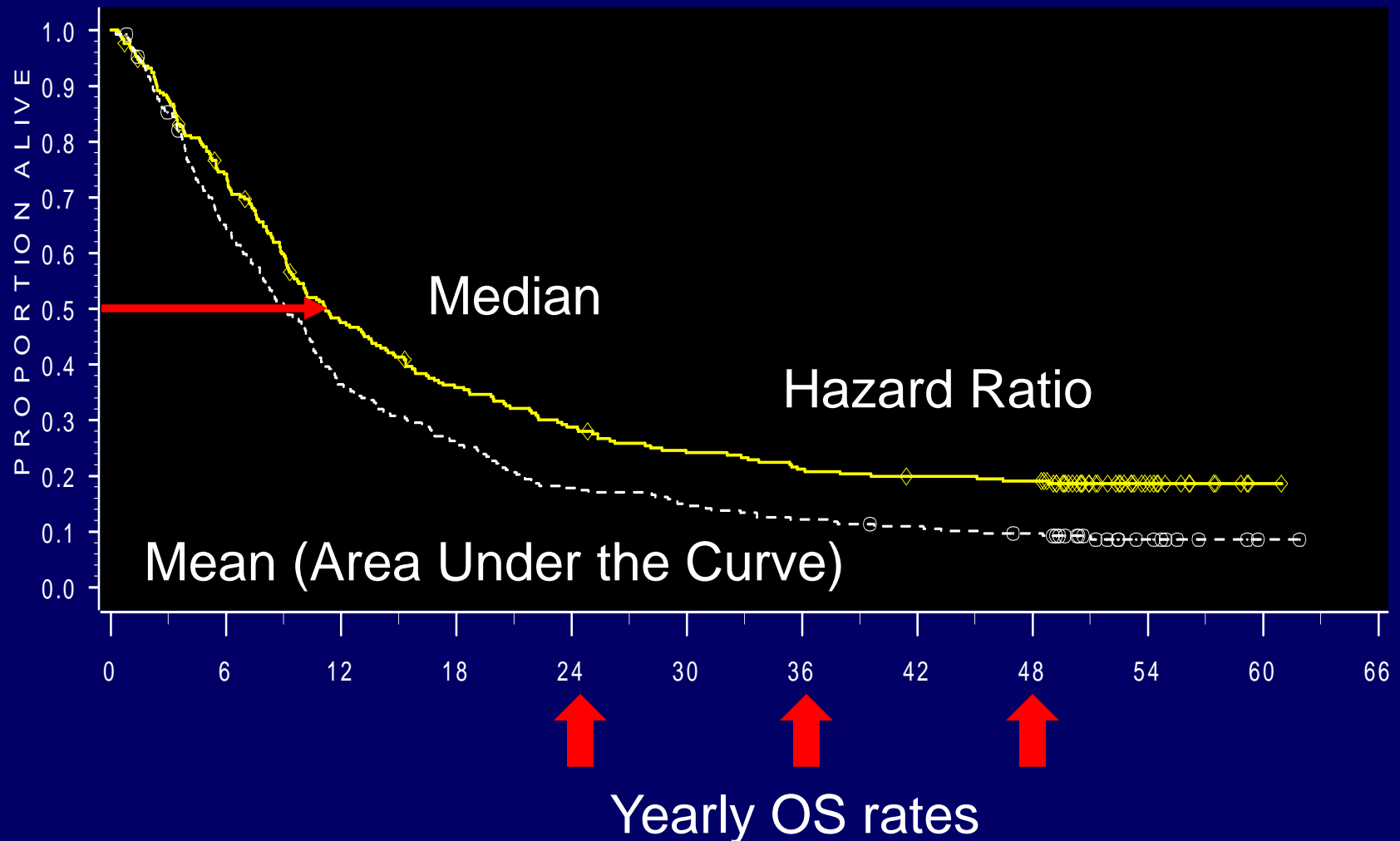
Additional Analysis Considerations

Prediction of timing of analyses

- ◆ Does the long-term survival alter projected study duration?

Additional Analysis Considerations

Summary Measures



Statistical Analysis Considerations

Primary analysis

- ◆ Remains log-rank test and Cox model?

Long-term survival

- ◆ Regulatory: Median vs. OS rates
- ◆ Market access: Mean
- ◆ Cure rate models

Delayed clinical effect

- ◆ Fleming-Harrington weighted log-rank test

Summary

- ◆ **Understand disease characteristics and MOA of therapy**
 - **Delayed clinical effect**
 - **Long-term survival**
- ◆ **Implications on study design and analyses**

Reference

- ◆ **Statistical issues and challenges in immuno-oncology, Tai-Tsang Chen, Journal for ImmunoTherapy of Cancer 2013, 1:18**