

Imaging as a predictor of therapeutic response

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Acknowledgement

- David Mankoff, MD, PhD

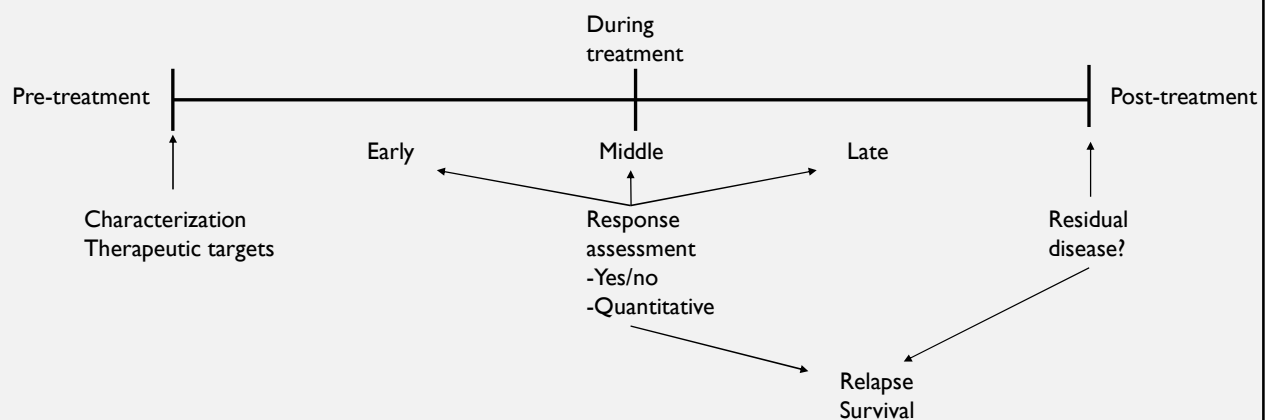
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Learning objectives

- List potential response biomarker imaging applications
- Describe the difference between prognostic and predictive biomarkers
- Discuss the approach to clinical trials designed to test the accuracy of imaging response biomarkers

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Imaging to guide cancer therapy



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How can biomarkers guide cancer therapy?

- Goals in cancer treatment
 - Characterize tumor biology pre-Rx
 - Individualized, specific therapy
 - Static response may be acceptable
- The implied needs for cancer biomarkers
 - Characterize tumor biology
 - Identify targets, predict response
 - Measure tumor response (early!)
 - Relate response to survival

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How can biomarkers guide cancer therapy?

- Goals in cancer treatment
 - Characterize tumor biology pre-Rx
 - Individualized, specific therapy
 - Static response may be acceptable
- The implied needs for cancer biomarkers
 - Characterize tumor biology – **Prognosis**
 - Identify targets, predict response – **Prediction**
 - Measure tumor response (early!) – **Response**
 - Relate response to survival – **Biologic response**

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Guidelines for biomarker studies: REMARK

COMMENTARY

Reporting Recommendations for Tumor Marker Prognostic Studies (REMARK)

Lisa M. McShane, Douglas G. Altman, Willi Sauerbrei, Sheila E. Tambo, Massimo Gion, Gary M. Clark for the Statistics Subcommittee of the NCI-EORTC Working Group on Cancer Diagnostics

1180 COMMENTARY

Journal of the National Cancer Institute, Vol. 97, No. 16, August 17, 2005

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EDITORIAL

Standards for Reporting Prognostic Tumor Marker Studies

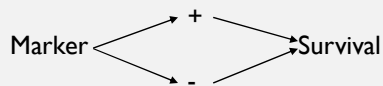
Todd A. Alonzo, Division of Biostatistics, University of Southern California Keck School of Medicine, Los Angeles, CA

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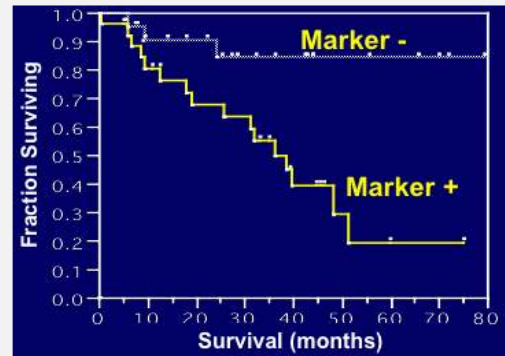
Imaging and therapeutic response: Prognosis

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Study design for prognosis

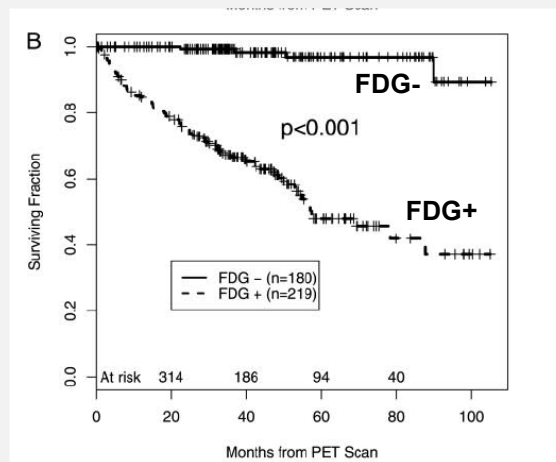
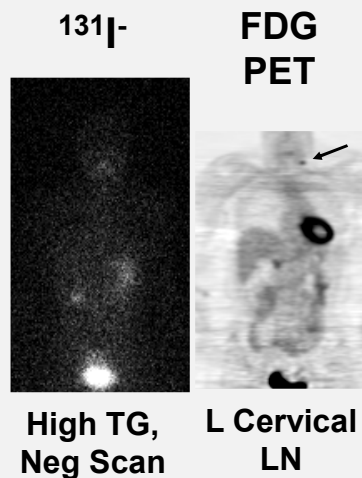


- In vitro examples:
 - Proliferation – Ki67
 - Receptor expression – ER
 - Oncogene expression – HER2



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FDG predicts survival in recurrent thyroid cancer - Robbins, JCEM, 2006



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Imaging hypoxia as the accumulation of a radiopharmaceutical



KA Krohn University of Washington

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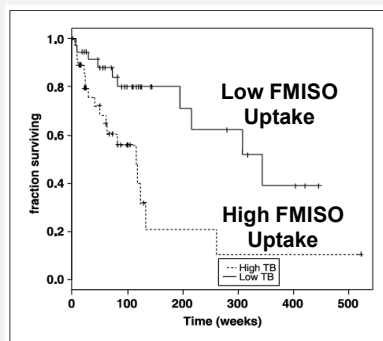
Tumor hypoxia quantified by PET predicts survival

**FMISO PET
Brain Tumor**



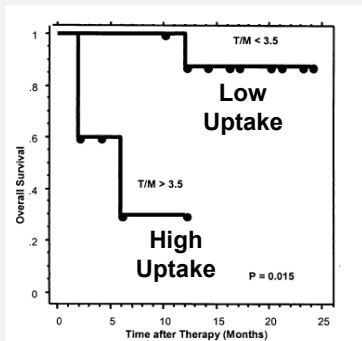
(Spence, Clin Cancer Res, 2008)

**FMISO PET
H & N Cancer**



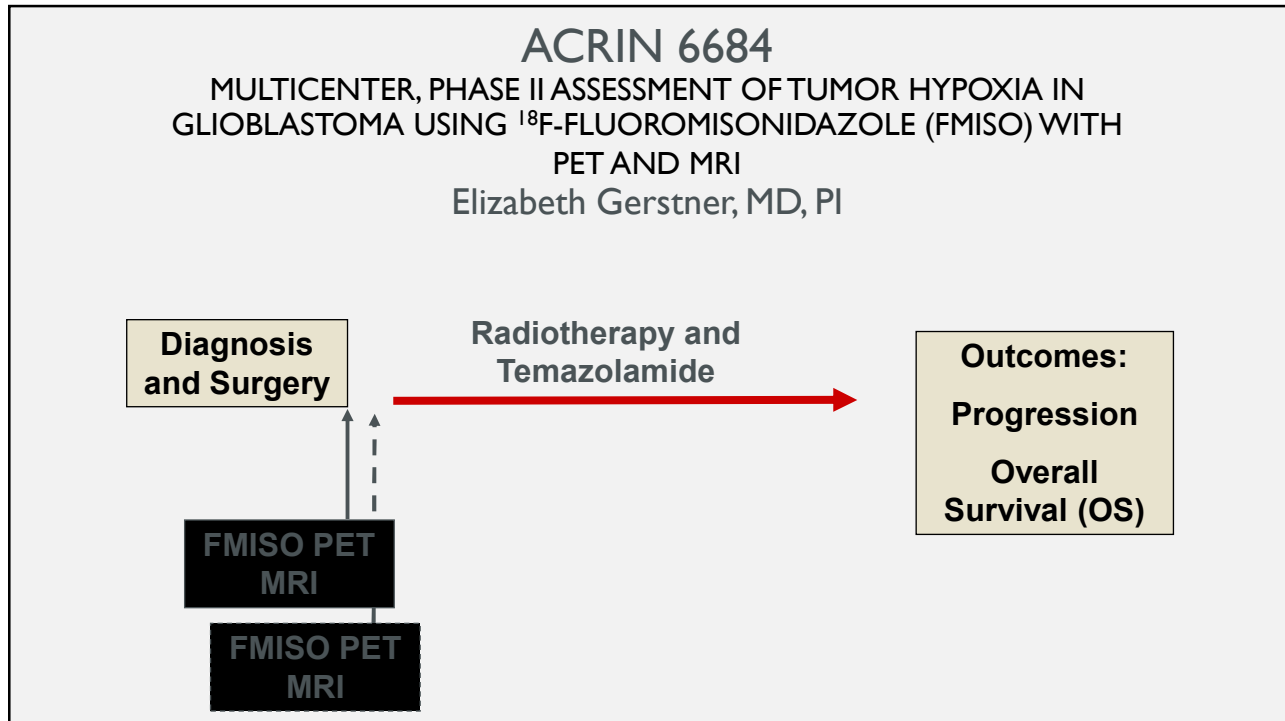
(Rajendran, Clin Can Res, 2007)

**Cu-ATSM PET
Cervical Cancer**



(Dehdashti, Int J Radiat Oncol Biol Phys, 2003)

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ACIN 6684: Hypoxia PET and MRI predict GBM PFS and OS

Table 3: Cox Regression Model

Parameter	Overall Survival Time			Progression Free Survival		
	Hazard Ratio	95% CI	p-value	Hazard Ratio	95% CI	p-value
<i>Univariate Model</i>						
SUVpeak	1.54	1.03, 2.38	0.045*	1.24	0.80, 1.91	0.35
FBmax	1.18	0.75, 1.81	0.50	0.93	0.61, 1.40	0.72
IVr	1.00	0.97, 1.03	0.90	1.01	0.98, 1.04	0.38
Mean K ^{trans}	1.17	1.02, 1.34	0.024*	1.10	0.99, 1.23	0.074
Median K ^{trans}	1.32	1.01, 1.72	0.045*	1.30	1.04, 1.63	0.021*
nCBV	1.11	0.99, 1.27	0.11	1.29	1.08, 1.54	0.0060*
nCBF	1.07	0.88, 1.29	0.51	1.18	1.01, 1.38	0.038*

Gerstner, Clin Cancer Res, 22:5079, 2016

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Imaging and therapeutic response: Prediction

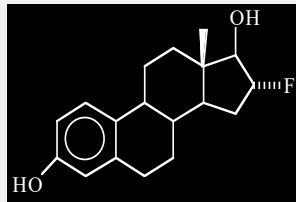
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Outcomes for cancer imaging: Prediction

- Predictor or response to specific therapy:
 - Positive – predicts who will respond
 - Negative – predicts who will **not** respond

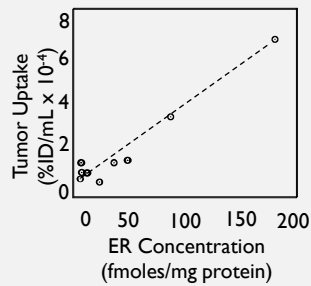
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¹⁸F-Fluoroestradiol (FES): PET Estrogen Receptor (ER) Imaging Provides a Quantitative Estimate of ER Expression



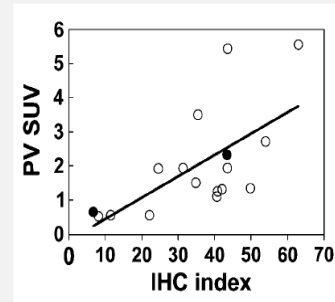
(Kieswetter, J Nucl Med, 25: 1212, 1984)

vs Radioligand Binding



(Mintun, Radiology 169:45, 1988)

vs IHC



(Peterson, J Nucl Med 49: 367, 2008)

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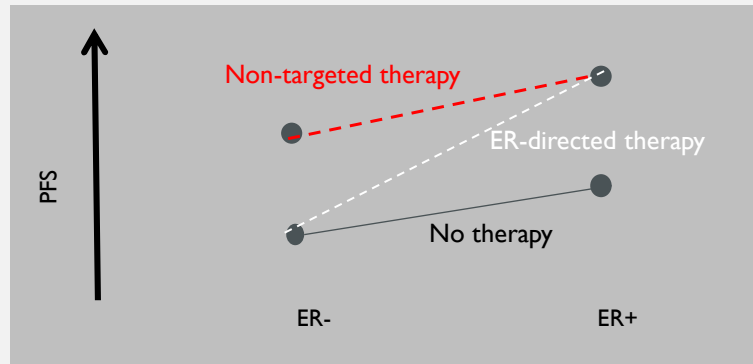
FES uptake predicts breast cancer response to hormonal therapy

	Pre-Rx	Post-Rx	
<p>Example 1</p> <ul style="list-style-type: none"> • Recurrent sternal lesion • ER+ primary • Recurrent Dz strongly FES+ 	<p>FES</p>	<p>FDG</p>	<p>Excellent response after 6 wks Letrozole</p>
<p>Example 2</p> <ul style="list-style-type: none"> • Newly Dx' d met breast CA • ER+ primary • FES-negative bone mets 	<p>FES</p>	<p>FDG</p>	

Linden, J Clin Onc, 2006

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Cancer markers: Prognostic, predictive, or both?



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Imaging and therapeutic response: Response

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Outcomes for cancer imaging: Response

- Accuracy of response assessment
 - Response or not - R versus NR
 - Degree of response – residual disease versus CR
- Surrogate outcome measure
 - Predictor of DFS, OS

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Categories of response

ANATOMIC

- Complete response
- Partial response
- Stable disease
- Progressive disease

FUNCTIONAL

- Complete metabolic response
- Partial metabolic response
- Stable metabolic disease
- Progressive metabolic disease

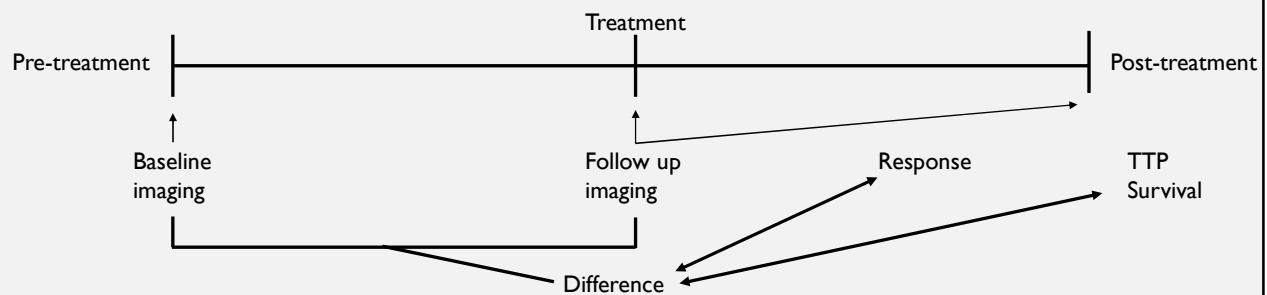
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Pseudoprogression

- The scan looks worse, but the patient is responding to the treatment
- Very difficult or impossible to differentiate from true progression
 - Usually, only time will tell

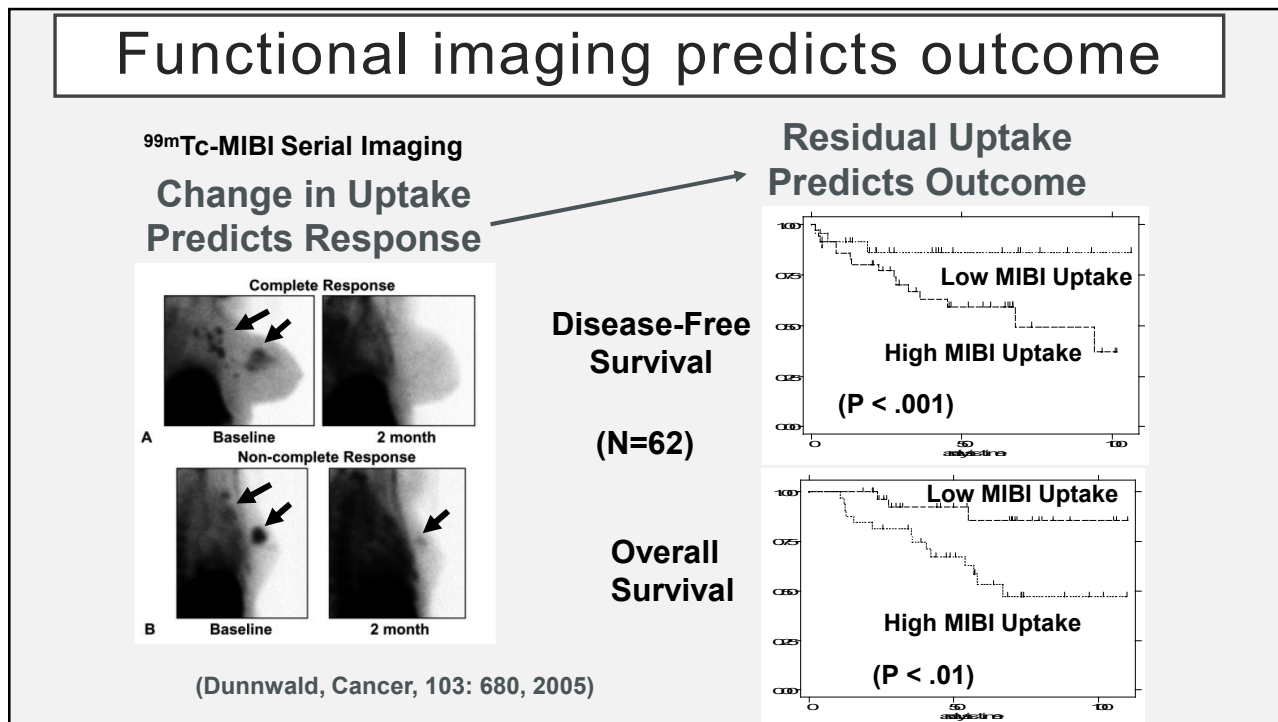
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Study design for response

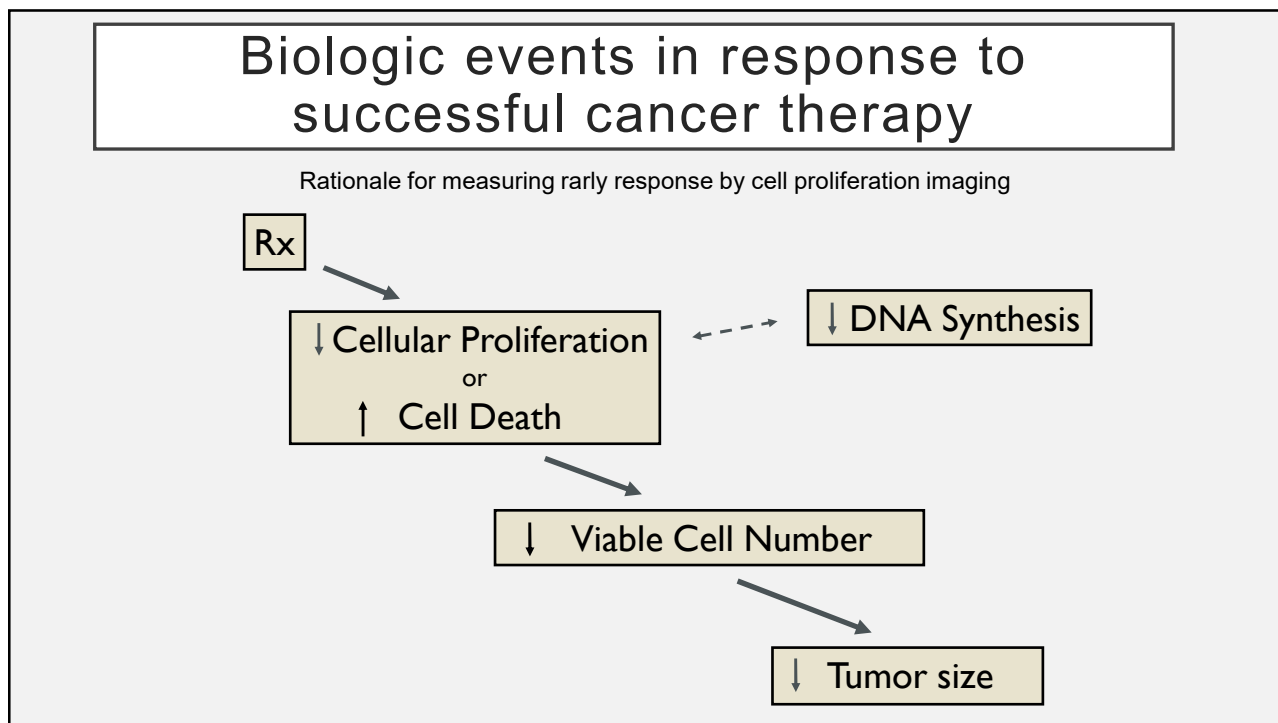


- Outcomes:
 - Sensitivity, specificity, ROC AUC for response
 - Predictor of TTP, Survival

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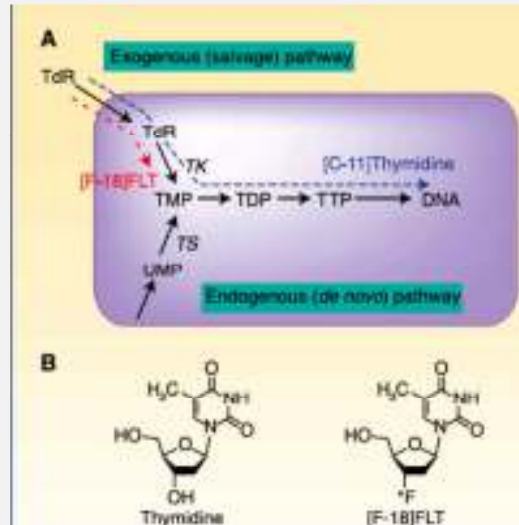
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Thymidine incorporation pathways

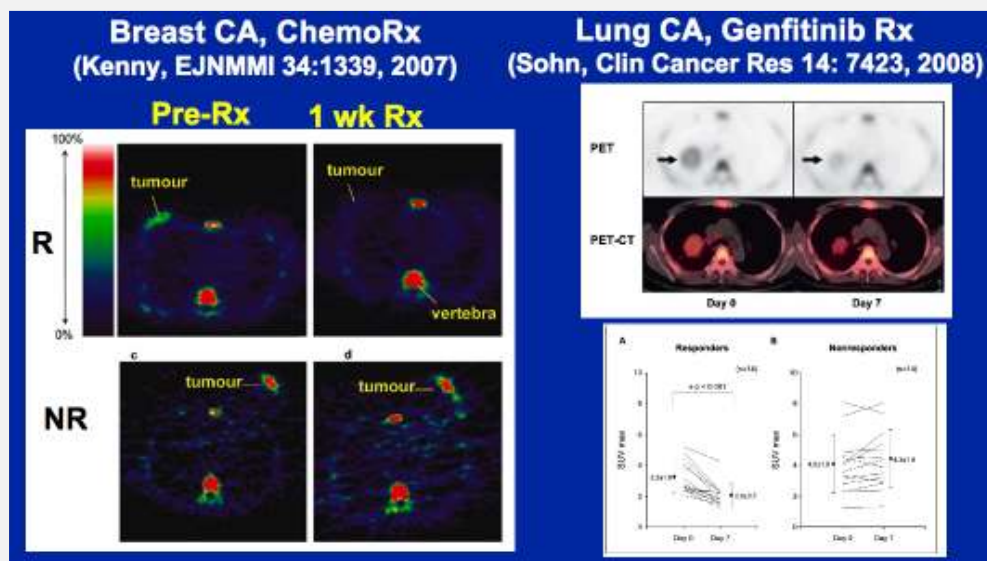
Imaging tumor proliferation



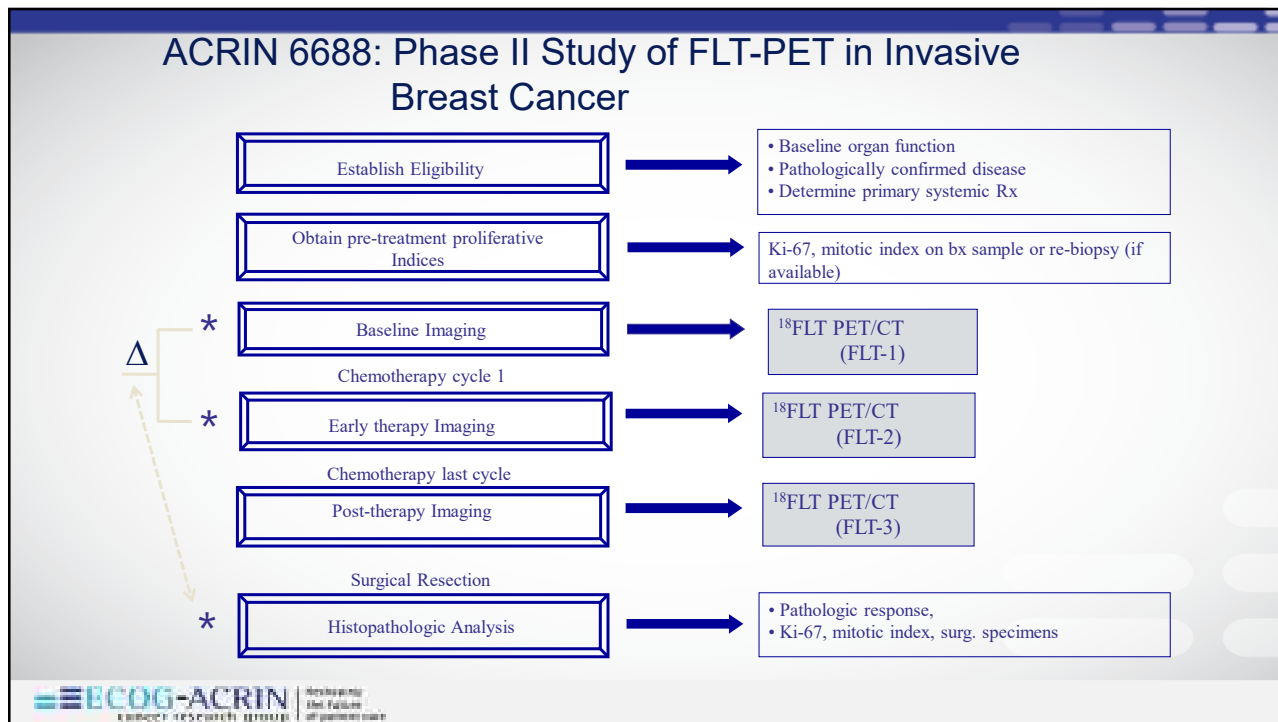
Mankoff and Eary, Clin Cancer Res 14:7159, 2008

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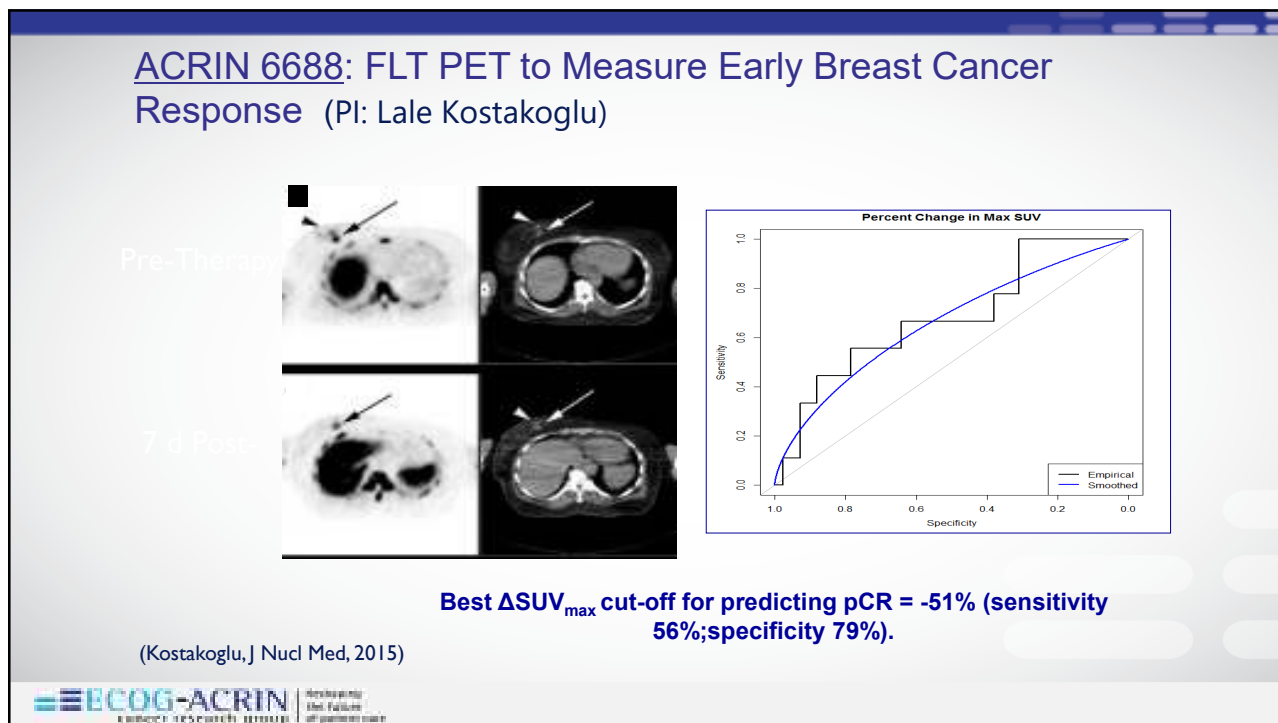
Early response measured by ^{18}F -fluorothymidine (FLT) PET/CT



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Imaging and therapeutic response:
Biologic response

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Outcomes for cancer imaging:
Biologic response

- Can functional/molecular response better predict outcome?
 - Predict DFS, OS, etc.
 - And what are the biologic insights
- Surrogate outcome measures?

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Goals of anti-cancer therapy

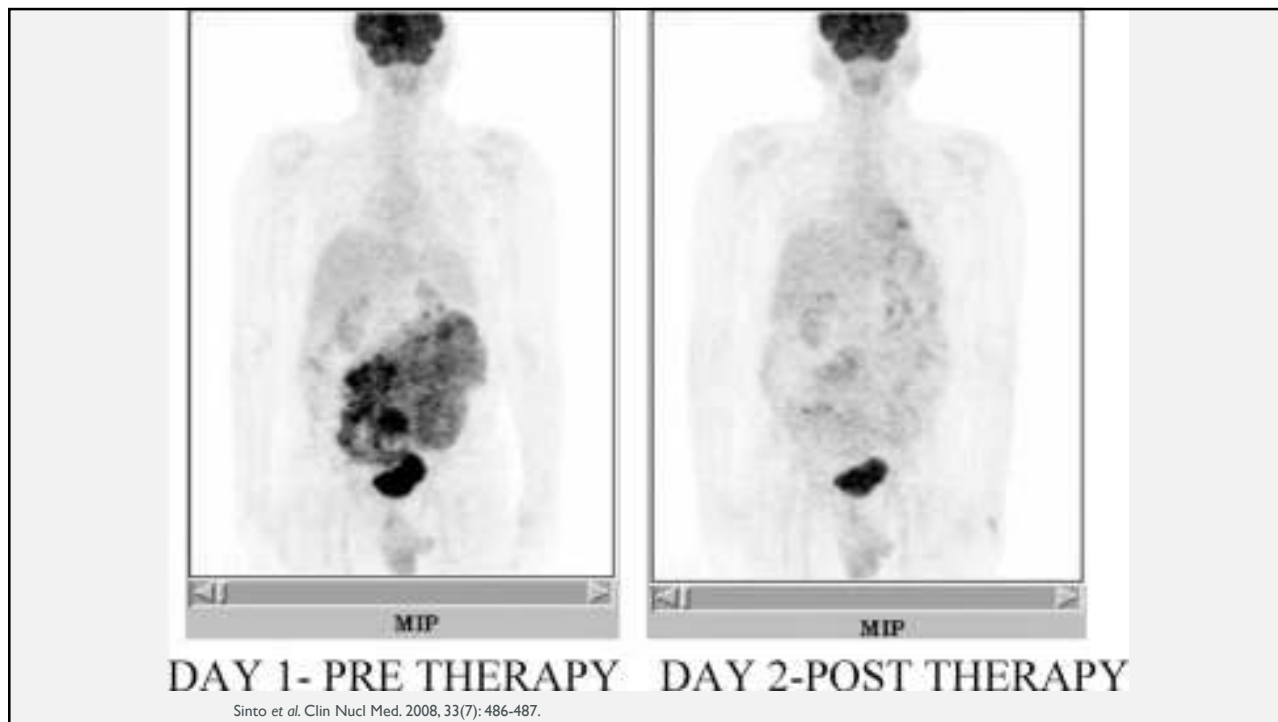
- Improve quality of life
- Improve duration of life
- Shrinking tumors in isolation not a goal of therapy
 - Anatomic response not a strong surrogate for clinical endpoints

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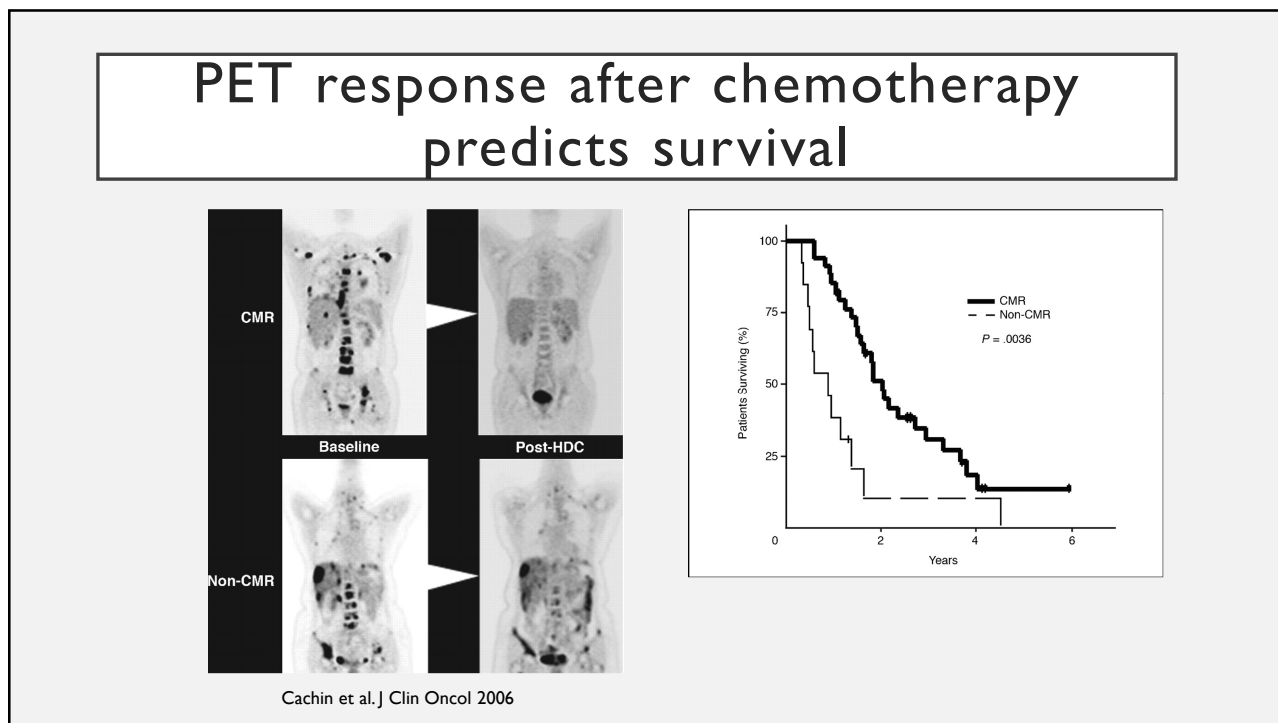
Functional imaging for response assessment

- Functional changes precede anatomic changes
- Functional changes can exist in the absence of anatomic changes
 - Cytostatic therapies
- More examples of functional imaging as predictive of clinical endpoints

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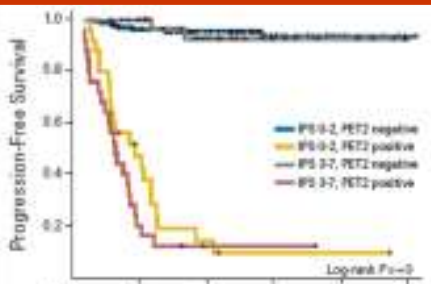
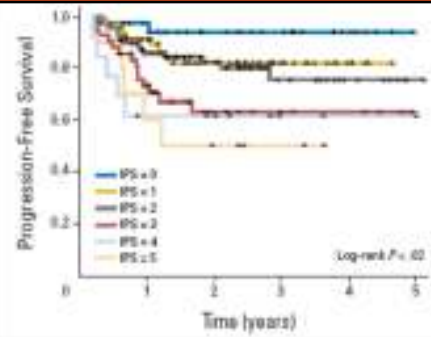


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Advanced Stage HL

- 260 HL patients, prospective
 - unfavorable stage IIA 26%
 - stage IIB 27%
 - stage III-IVB 47%
- End-point: 2yr PFS, med f/u 2.2 y
- 79% CR; 16% prog <6mo; 4% relapse
 - PPV 86%
 - NPV 95%
 - Sens and spec: 81% and 97%
 - 2-yr PFS for PET2- vs PET2+ 95% vs 13%,
Positive PET definition uptake > MBP

(courtesy of Lale Kostakoglu)



PET-2 was significant overshadowing the prognostic value of IPS

Gallamini et al. J Clin Oncol. 2007 ;25:3746

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Post-therapy FDG PET predicts survival in lymphoma Zanoni, Q J Nucl Med Mol Imag 55:633, 2011

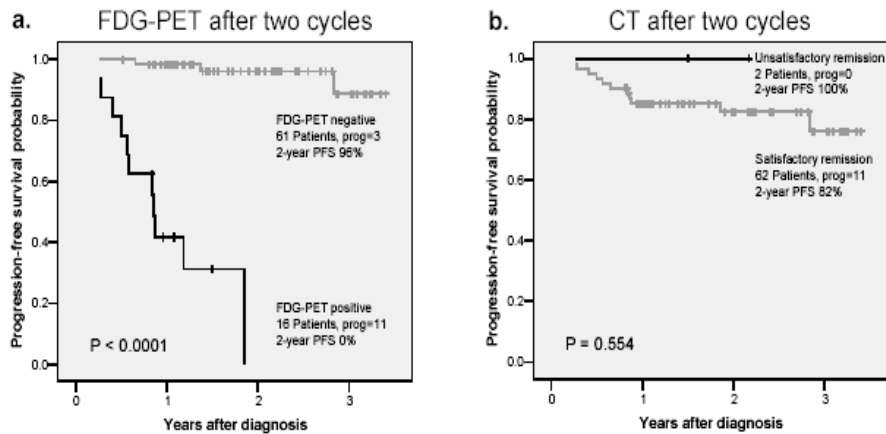
TABLE 1.—Prognostic value of post-treatment 18F-FDG PET for Aggressive NHL and HL

AUTHOR	Patients (n)	DISEASE	NOTES	FOLLOW-UP (months)	PFS PET+ vs PET-
Zinzani 1999 18	44	DLBCL/101	Abdominal mass in all	20	1 yr: 15% vs 95% 2 yr: 0% vs 99%
Jerusalem 1999 11	54	LH/LNH	RM in 24	21	1 yr: 0% vs 86%
Mansy 2001 11	24	LH/LNH	RM in 24	29	50% vs 75%
Mikhaeel 2000 13	45	Aggressive LNH	RM in 17	30	1 yr: 0% vs 84%
Spangon 2000 14	93	Aggressive LNH	RM in 24	22	2 yr: 4% vs 87%
Jurevic 2002 15	38	Aggressive LNH	-	15.5	1 yr: 8% vs 88%
de Wit 2001 16	37	LH	RM in 37	25.6	54% vs 96%
Weischenb 2001 17	28	LH	RM in 28	28	1 yr: 40% vs 99%
Spangon 2001 16	60	LH	RM in 45	31	2 yr: 4% vs 85%
Mikhaeel 2002 13	65	LH	-	36	1 yr: 0% vs 93%

Abbreviations: RM, residual mass; PET+, positive; PET-, negative

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Early interim FDG-PET and prognosis



(courtesy of A Shields, Karmanos Cancer Center)

M Hutchings, Blood, 2006

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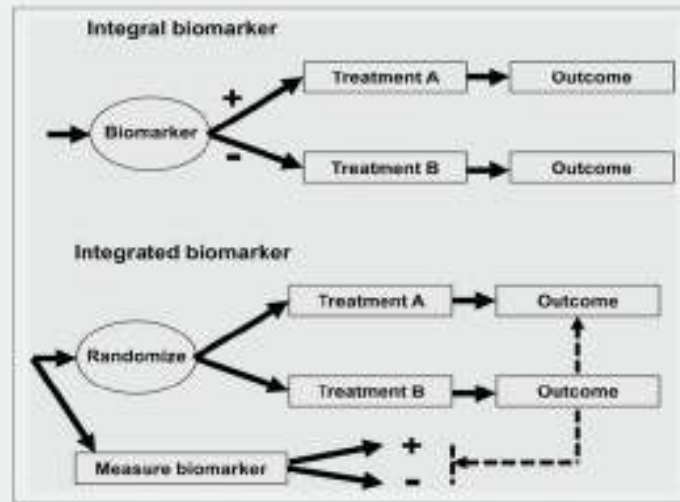
Example: Rectal cancer neoadjuvant therapy

- Meta-analysis of 34 studies
 - Most studies showed FDG PET/CT predictive of pathologic response to neoadjuvant therapy
- Pooled cohort of 1526 patients
- Pooled response cutoff of 63%

Maffione et al., AJR 2015

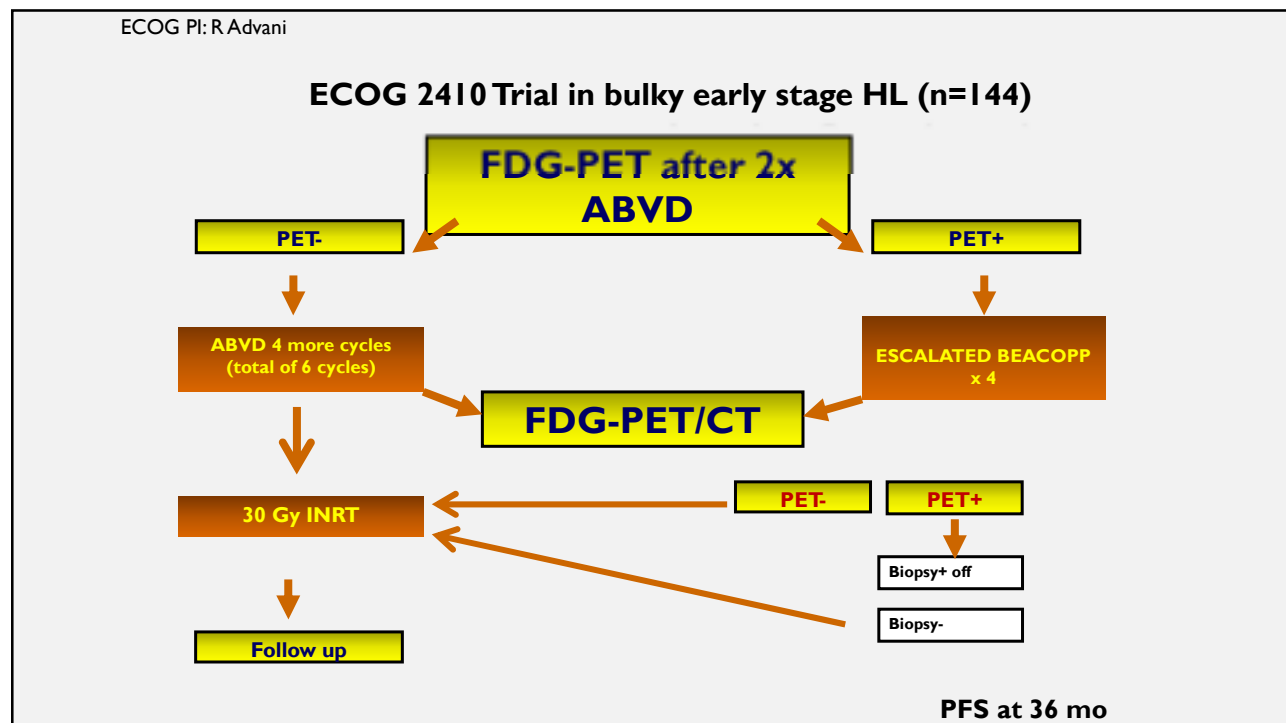
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Imaging biomarker in cancer trials: Integrated vs integral markers



(Mankoff, J Nucl Med 55:525, 2014)

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Imaging as a biomarker: Summary

- Imaging to guide treatment – imaging as a disease biomarker
 - Prognosis – How aggressive is the disease?
 - Prediction - Will the treatment work?
 - Response - Is the treatment working?
 - Biologic response
 - Can response predict survival?
 - Can we use insights from imaging to adapt therapy?