Radiation Therapy and Genomic Interactions in Breast Cancer Patients

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Radiation Therapy & Genomic Interactions in Breast Cancer Patients.

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Radiation Therapy in Oncology

Why talk about RT?

- It is the most commonly utilized single therapeutic agent in oncology (up to 60% of all cancer patients)
- Responsible for 40% of all cancer cures
- Highly cost effective



The empiric basis of modern radiotherapy

1911



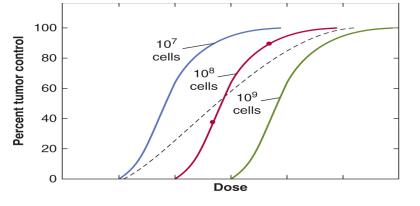
Claudius Regaud – Fractionated RT more effective than single dose

1920-30

THE RESULTS AND METHODS OF TREATMENT OF CANCER BY RADIATION

PROFESSOR HENRI COUTARD, M.D.
PARIS. FRANCE

1940-60



Coutard reports fractionated RT cures head and neck cancer

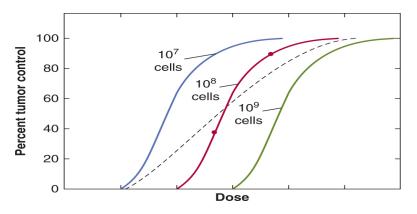
Fletcher summarizes required doses for optimized tumor control in head and neck

50 Gy – Subclinical disease

60 Gy - Microscopic disease

70 Gy – Macroscopic disease

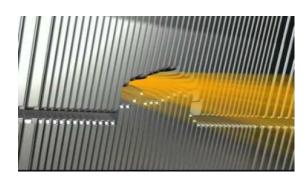
The empiric basis of radiotherapy



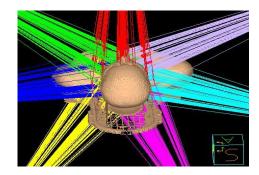
- Radiation damage is probabilistic
- Assumption tumors are homogenous
- Everyone has the same opportunity to benefit

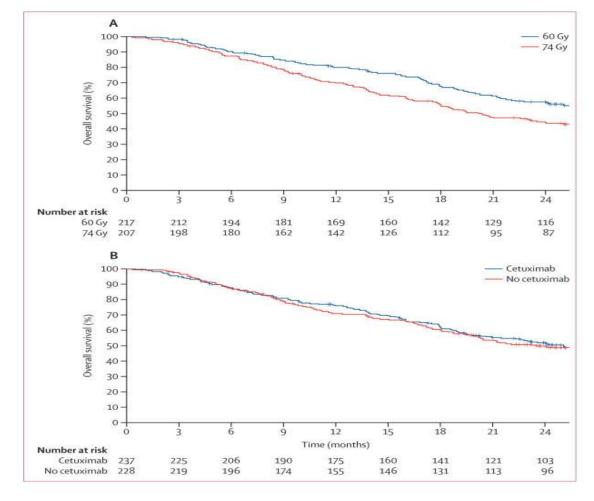
Dogma – "It is axiomatic that a higher dose results in a higher effect"

~Perez and Brady – Textbook of Radiation Oncology



Simple idea: If we increase the total dose we will cure more patients





RTOG 06-17

Randomized over 400 pts 60 Gy vs. 74 Gy 60 Gy was superior to 74

Other negative studies:

RTOG 0126

70.2 vs. 79.2 No difference in 10-yr OS

RTOG (Esophagus)

50.4 vs. 64.8 No difference in OS

Bradley J et al. Lancet Oncol. 2015 Feb; 16(2): 187–199.

Modern oncology is genomics-based

- Medical Oncology
- Field has transitioned from an empiric basis to a scientific basis
- Biology guides decisions

Oncotype DX – no chemotherapy breast HER-2 neu - Herceptin ALK fusion gene – crizotinib Sequencing – Personalized Genomic Reports

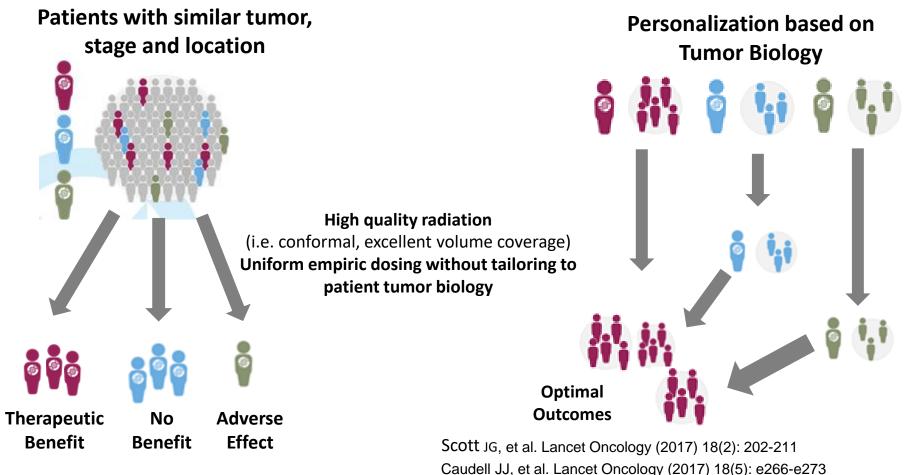
Radiation Oncology is still mostly empiric







A Need for a Personalized Approach in Radiation Oncology



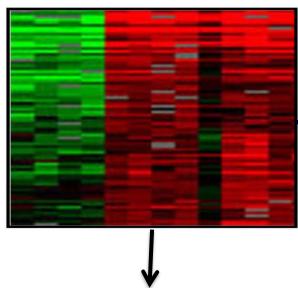
Development of the Radiosensitivity Index (RSI)

Validated Surviving Fraction at 2 Gy

	Recorded		Recorded	
Cell Line	SF2	Cell Line	SF2	
BREAST_HS578T	0.79	COLON_COLO205	0.69	
BREAST_MDAMB231	0.82	COLON_HCC-2998	0.44	
COLON_HCT116	0.38	COLON_HT29	0.79	
COLON_HCT15	0.4	COLON_KM12	0.42	
COLON_SW620	0.62	MELAN_LOXIMVI	0.68	
LEUK_CCRFCEM	0.185	MELAN_M14	0.42	
LEUK_HL60	0.315	MELAN_MALME3M	0.8	
LEUK_MOLT4	0.05	MELAN_SKMEL28	0.74	
MELAN_SKMEL2	0.66	MELAN_SKMEL5	0.72	
NSCLC_A549ATCC	0.61	MELAN_UACC257	0.48	
NSCLC_H460	0.84	MELAN_UACC62	0.52	•
NSCLC_HOP62	0.164	NSCLC_EKVX	0.7	
NSCLC_NCIH23	0.086	NSCLC_HOP92	0.43	
OVAR_OVCAR5	0.408	OVAR_OVCAR3	0.55	
RENAL_SN12C	0.62	OVAR_OVCAR4	0.29	
BREAST_BT549	0.632	OVAR_OVCAR8	0.6	
BREAST_MCF7	0.576	OVAR_SKOV3	0.9	
BREAST_MDAMB435	0.1795	PROSTATE_DU145	0.52	
BREAST_T47D	0.52	PROSTATE_PC3	0.484	
CNS_SF268	0.45	RENAL_7860	0.66	
CNS_SF539	0.82	RENAL_A498	0.61	
CNS_SNB19	0.43	RENAL_ACHN	0.72	
CNS_SNB75	0.55	RENAL_CAKI1	0.37	
CNS_U251	0.57	RENAL_UO31	0.62	

Gene mutation	Wild Type	Mutant	Total
RAS (H,N, or K)	33	15	48
TP53	17	31	48

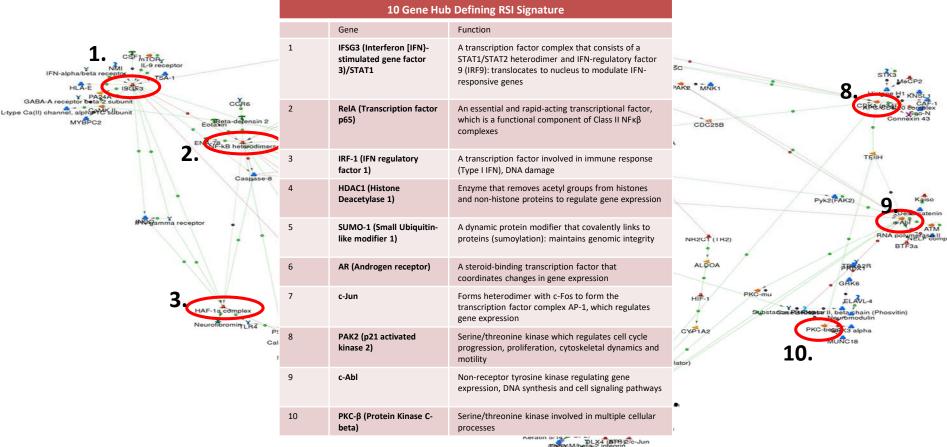
Basal transcriptome of cell lines



 $\begin{array}{l} {\rm SF2_x} = k_0 + k_1(y_x) + k_2({\rm TO}) + k_3({\rm ras\ status}) + k_4({\rm p53} \\ {\rm status}) + k_5(y_x)({\rm TO}) + k_6(y_x)({\rm ras\ status}) + k_7({\rm TO})({\rm ras\ status}) \\ + k_8(y_x)({\rm p53\ status}) + k_9 \ ({\rm TO})({\rm p53}) + k_{10} \ ({\rm ras\ status})({\rm p53\ status}) \\ + k_{12}(y_x)({\rm ras\ status})({\rm p53\ status}) \\ + k_{14} \ (y_x)({\rm TO})({\rm ras\ status})({\rm p53}) \\ \dots \end{array}$

(Eschrich et al. IJROBP 2009)

500 Gene Radiosensitivity Network



500 gene network identified by least sum squares approach Network hubs identified by having > 5 connections

(Eschrich et al. IJROBP 2009)

Next, the expression levels of the 10 hub genes were ranked (greatest to least) and modeled with regression analysis to predict SF₂ in each of the 48 cell lines

```
RSI = -0.0098009*AR + 0.0128283*cJun + 0.0254552*STAT1- 0.0017589*PKC -
```

0.0038171*RelA + 0.1070213*cABL - 0.0002509*SUMO1 - 0.0092431*PAK2 -

0.0204469*HDAC1 -0.0441683* IRF1

RSI provides a continuous score from 0-1 Lower the score= radiosensitive

Clinical Validation of RSI

Radiation Treatment N=1442

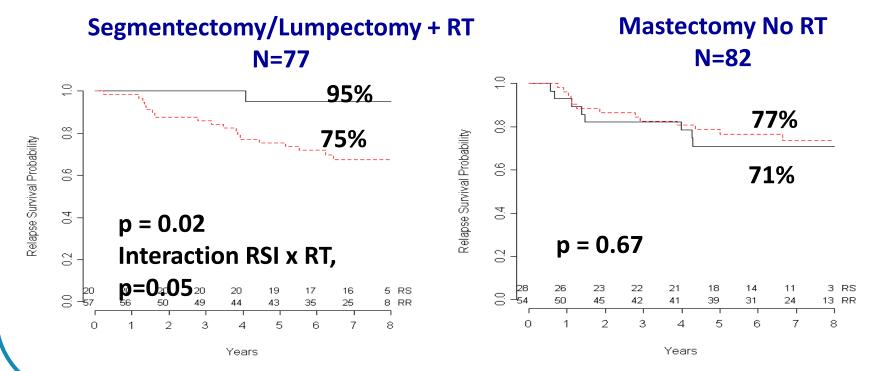
			Hazard Ratio	Clinical Outcome	
Disease Site	N	Endpoint	(Ref. Radioresistance)	RS vs RR	p-value
Breast (Karolinska)	77	RFS	0.13 (0.02-1.0)	95% vs. 75% (5-yr)	0.02
Breast (Erasmus)	288	DMFS	0.57 (0.33-0.98)	77% vs. 64% (5-yr)	0.04
Breast (Curie, NKI	343	LRFS	0.23 (0.1-0.53)		0.0006
Lung (Moffitt)	53	DFS	0.42 (0.25-0.92)	63% vs. 22% (5-yr)	0.02
Lung (Dir Chall)	27	DFS	0.44 (0.16-1.18)		0.09
Lung (Korea)	16	DFS	0.27 (0.03-2.17)	75% vs. 25% (5-yr)	0.18
GBM (TCGA)	214	os	0.57 (0.38-0.85)	<u></u>	0.005
Pancreas (Moffitt)	48	os	0.42 (0.19-0.94)		0.04
Prostate (Mayo)	82	DMFS		94% vs. 72% (10-yr)	0.03
Prostate (TJU)	132	BFFS		80% vs. 60% (5-yr)	0.026
Head and Neck (NKI)	92	LRFS		86% vs. 61% (2-yr)	0.05
Rectal	14	Response			0.03
Esophageal	12	Response			0.05
Melanoma	11	OS	0.09 (0.01-0.81)	100% vs. 14.3% (2-yr)	0.009
Liver metastasis	33	LC (s/p SBRT)	•	100% vs. 59% (2-yr)	0.019

No Radiation Treatment N=877

Disease Site	N	Endpoint	Hazard Ratio (Ref. Radioresistance)	Clinical Outcome RS vs RR	p- value
Breast (Karolinksa)	82	RFS	1.21 (0.50-2.91)	77% vs. 71% (5-yr)	0.67
Breast (Erasmus)	62	DMFS	1.06 (0.23-4.83)	80% vs. 81% (5-yr)	0.94
Lung (Moffitt)	42	RFS	1.09 (0.45-2.65)		0.98
Lung (Dir Chall)	47	DFS	0.93 (0.50-1.79)	19% vs. 14% (5-yr)	0.84
GBM (TCGA)	52	os		5% vs. 5% (1-yr)	0.64
Pancreas (Moffitt)	25	os	0.76 (0.29-1.99)	69% vs. 67% (2-yr)	0.58
Prostate (Mayo)	536	DMFS		70% vs. 71% (10-yr)	0.58
Melanoma	31	os		91.7% vs. 63% (2-yr)	0.19

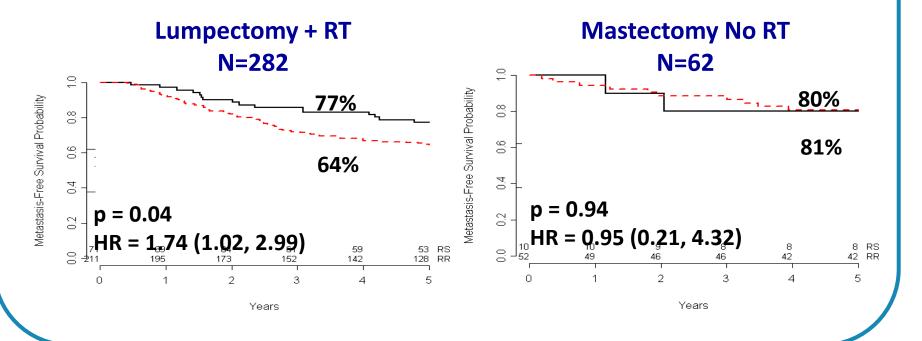
Abbreviations: RS, radiosensitive; RR, radioresistant; RFS, relapse-free survival; DMFS, distant metastasis-free survival; LRFS, local-regional relapse-free survival; DFS, disease-free survival; RFFS, biochemical failure-free survival; LC, local control; OS, overall survival

RSI: Predicts Outcome Only in RT-Treated Patients



Eschrich SA et al (2012) Clin Can Res 18:5134-43

RSI Predicts Distant Metastasis Risk Only in RT-Treated Patients

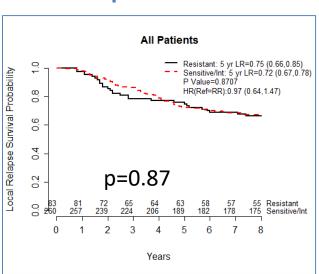


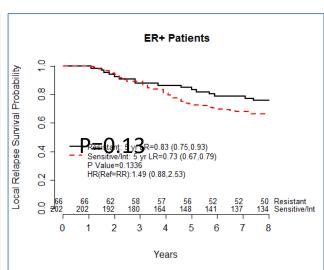
RSI predicts for local recurrence in breast cancer only in ER negative

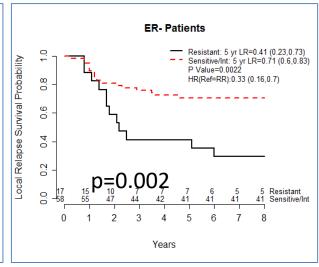
All patients

ER + Patients

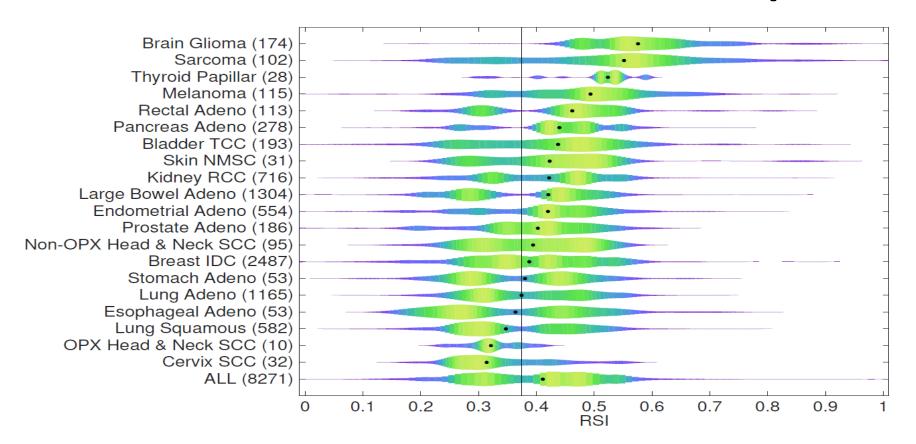
ER - Patients







RSI Distribution in 8,271 tissue samples



Grass, D et al (ASTRO 2017) (in preparation)

RSI: A Novel Assay of Tumor Radiosensitivity

RSI

- Biomarker of radiosensitivity
 - Developed based on SF2
- Clinically Validated
 - RT-specific, Disease-site independent
 - 13 independent datasets, 2,319 patients
 - Assessed in 13,343 patient samples



RSI: Integrating Genomics Into Clinical Practice

- RSI
 - RT benefit (response) is not uniform
 - RT Dosing is Uniform. Why?
- Physical Dose (same) vs. Tumor effect (different)
 - A given RT dose has varied effects on individual tumors
 - Uniform physical dose is biologically imprecise
- How about prescribing RT to tumor effect?
 - GARD Genomic Adjusted Radiation Dose
 - Dose Can be adjusted to account for biological heterogeneity

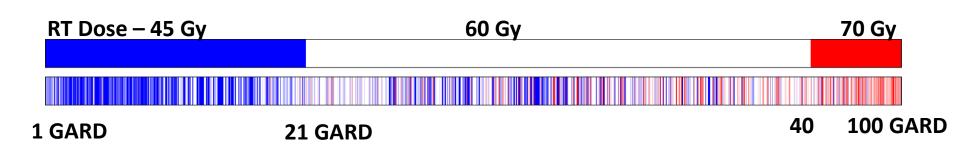


Deriving GARD by combining the LQ model and RSI

Deriving GAD from individual patient RSI (1) (2)(3)Genomically Adjusted Dose effect (GAD) Linear-Quadratic Patient specific Genomically Adjusted 2 Gy physical model of survival radiobiological Dose effect dose after n doses of d Gray parameters $= \frac{\ln RSI + \beta nd^2}{-nd}$ $SF = e^{-n(\alpha d + \beta d^2)}$ $GAD = nd(\alpha + \beta d)$ in experiments n & d trial specific, RSI ~ SF2 a patient specific, 0.2 0.4 (n=1 and d=2)β constant Radiation Sensitivity Index (RSI)



From Physical to Genomic RT Dose



Physical Dose – Discrete

N=8,271
Doses are assigned

GARD – Continuous

Higher Dose is not always associated with higher effect

Scott JG (2017) Lancet Oncology 18:202-11

GARD: Clinical Validation

	Median follow-up (months)	Events	Radiotherapy dose range (Gy)	GARD range	Endpoint	HR from multivariable analysis (95% CI)	p value	Adjustment factors
Erasmus Breast Cancer Cohort (n=263)	60	23	40-74	4-01-104-25	Distant- metastasis-free survival*	2·11 (1·1-3·9)	0.018	Oestrogen and progesterone receptor status, age, surgery (vs lumpectomy), and T-stage
Karolinska Breast Cancer Cohort (n=77)	87	19	50	8–60	Relapse-free survival†	7-4 (1-4-138)	0.014	Hormonal therapy, chemotherapy, and oestrogen and progesterone receptor status
Moffitt Lung Cancer Cohort (n=60)	37	23	45-70	15-125	Local control‡	3.4 (1.3-9.1)	0.016	Surgery, stage, histology, lymphovascular invasion
TCGA Glioblastoma Cohort (n=98)	11	76	12-6-97-0	0-4-46-0	Overall survival§	1.9 (1.1-3.3)	0.019	Age, performance status
Moffitt Pancreas Cancer Cohort (n=40)	68	27	45-54	16–40	Overall survival§	2.6 (1.1–6.0)	0.029	CA 19·9, margin lymph nodes

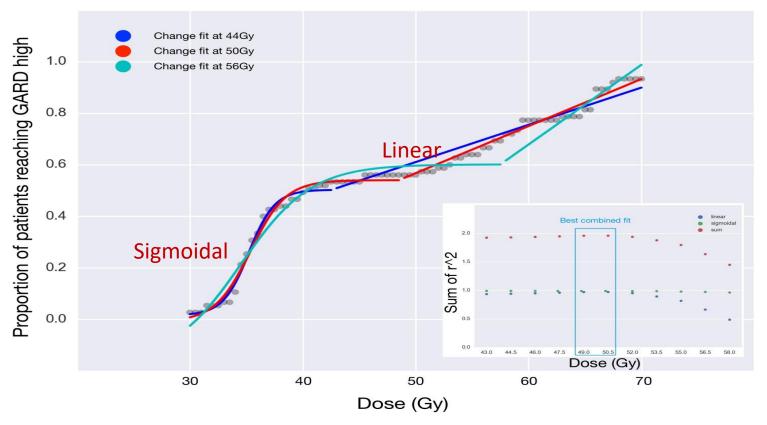
TCGA=The Cancer Genome Atlas. *Primary endpoint defined as distant recurrence in the first 5 years following completion of primary treatment. †Primary endpoint defined as any relapse distant, regional, or local from the end of primary treatment. ‡Defined in this study as time from surgical resection to cancer recurrence within the irradiated field. If no event occurred, then cases were censored at the date of last clinical evaluation. Cases in which more than 4 months elapsed without a clinical evaluation were censored at the date of antecedent clinical evaluation. 22 §Defined in this study as the interval from surgery to date of death.

Table 2: Clinical cohort description and multivariate analysis for the effect of GARD on selected endpoints

GARD predicts local control in ER negative breast cancer

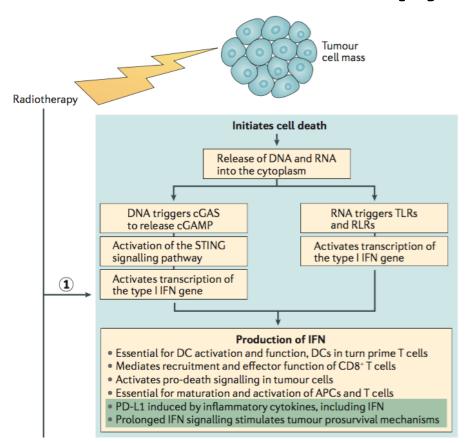
ER-				
		HR	95% CI	P-value
GARD		0.91	0.84-0.98	0.0082
Node status (ref: pN-)	pN+	2.02	0.93-4.55	0.08
LVSI (ref: No)	Yes	2.11	0.89-4.66	0.09
ER +				
Age		0.96	0.93-1.0	0.08
Grade (ref: I & II)	III	1.62	1.01-2.6	0.04
DCIS (ref: No DCIS)	DCIS	1.54	0.95-2.59	0.08
LVSI (ref: No)	Yes	1.25	0.76-2.0	0.37

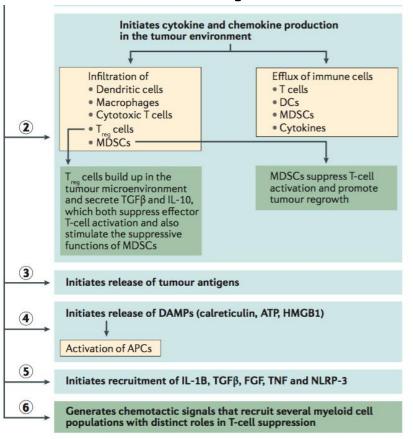
Personalized Genomic-Based RT Dose for ER negative breast cancer



Ahmed KA et al (2018) Submitted

Radiation Therapy & Immune Response





Experimental Approach:

- 10,469 primary macrodissected tumors were analyzed via an IRB-approved prospective de-identified tissue collection protocol at Moffitt since 2006 (Total Cancer Care)
 - Prospective pathology quality control (PQC) data of percent cellularity, stroma, malignancy & necrosis in each sample
 - All samples with gene expression data Affymetrix GeneChips (60,607 probe sets representing ~30,000 unique genes

Radiosensitivity (RSI)

- Samples stratified by median RSI score for each tumor type
- RSI low =radiosensitive
- RSI high = radioresistant

ESTIMATE (Yoshihara et al. Nature Comm. 2013)

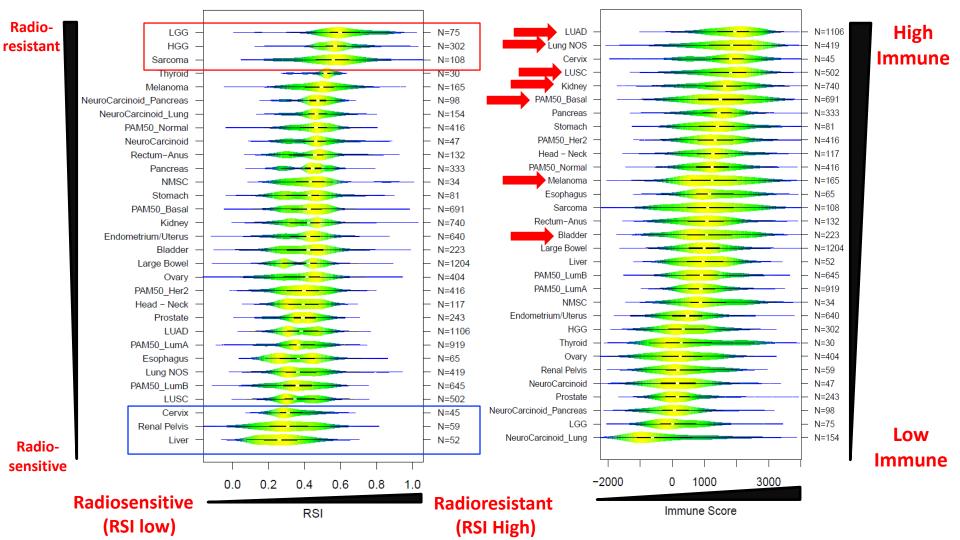
- Stromal cell presence
- Immune cell presence
- Tumor Purity

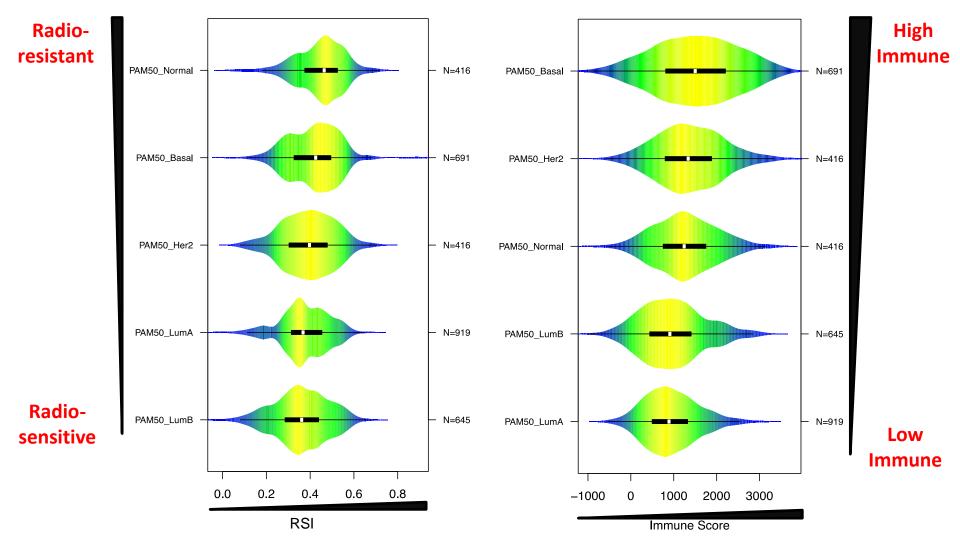
CIBERSORT (Newman et al. Nature Methods 2014)

 Relative presence of 22 distinct tumor-infiltrating lymphocytes (TILs) in complex tissue mixtures

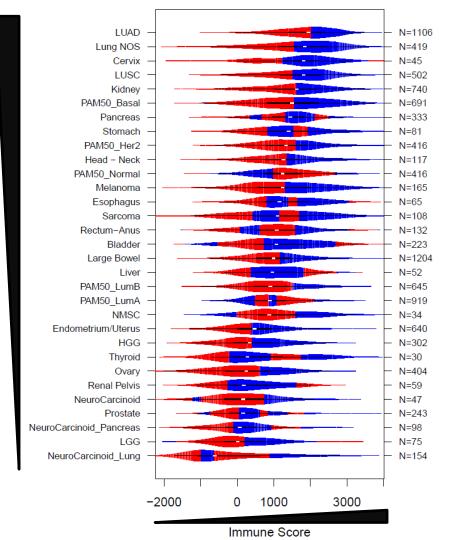
Total Cancer Care
PQC Data

Normalized cellular abundance of TILs





High Immune



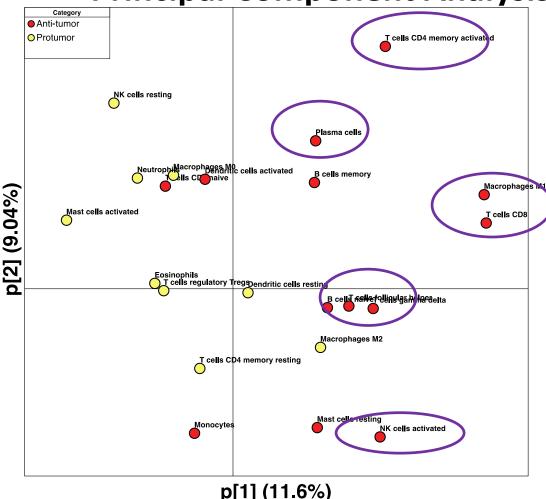
Radiosensitive= BLUE

Radioresistant = RED

For a given immune score there is an enrichment of radiosensitive tumors in samples also defined to have the highest immune cell presence, though with some heterogeneity

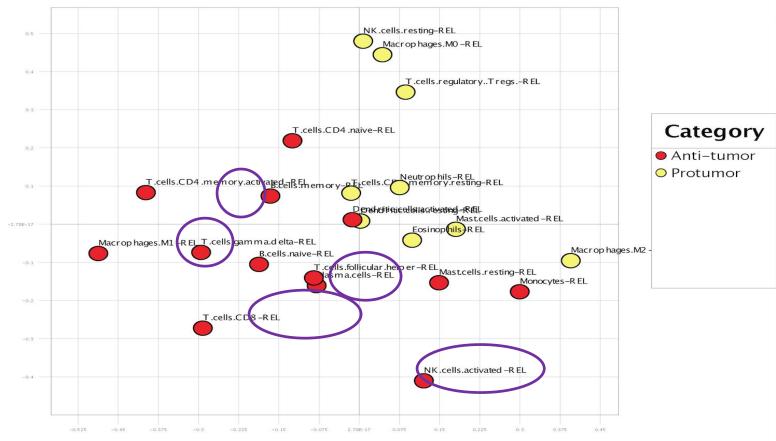
Low Immune

Principal Component Analysis and TIL Analysis



- Analysis of 10,469 samples across all tumor types
 - Combination of laboratory-based descriptions of TIL activity (color-coded) and unsupervised clustering demonstrates TIL groupings with similar biology independent of radiosensitivity
 - CD4⁺ memory activated T cells
 - Plasma cells
 - M1 Macrophages
 - CD8⁺ T cells
 - NK cells (activated)
 - T cell follicular helper

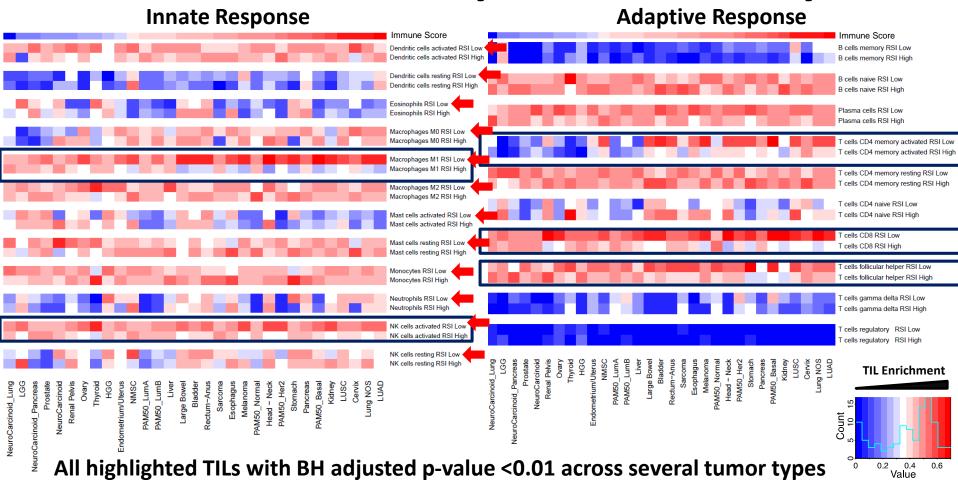
All Breast Cancer Types



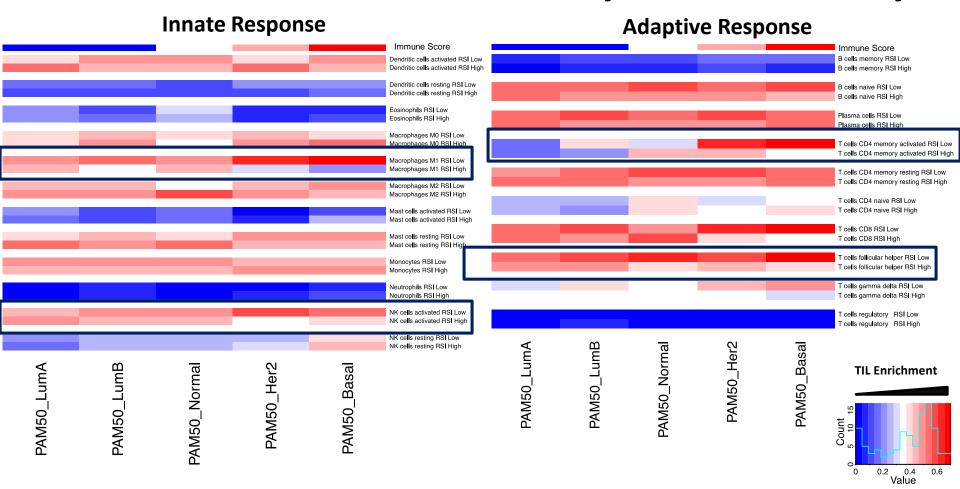
p[2] (9.43%)

p[1] (11.2%)

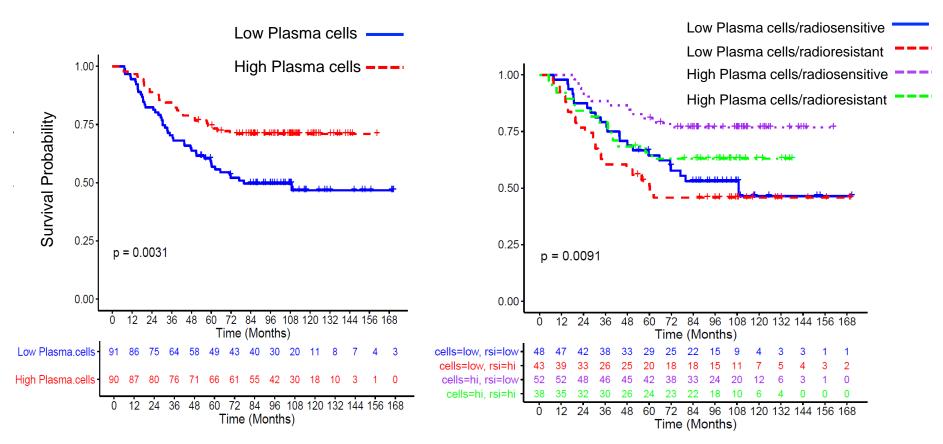
TIL Enrichment by Radiosensitivity



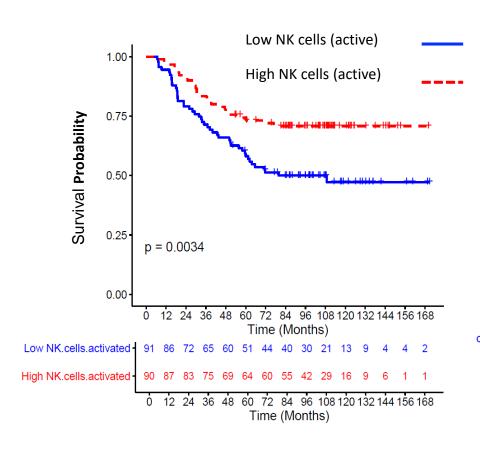
Breast Cancer TIL Enrichment by Radiosensitivity

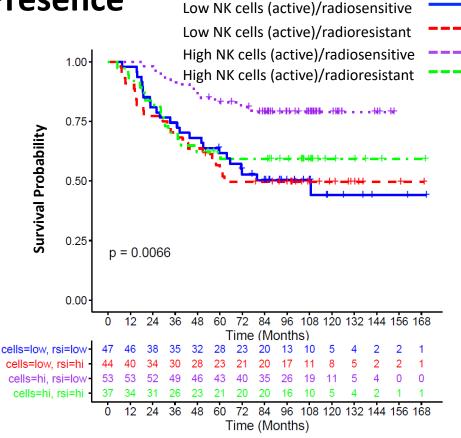


Integration of Intrinsic Radiosensitivity and TIL Presence



Integration of Intrinsic Radiosensitivity and TIL Presence





A New Framework for Radiation Oncology

- Conclusions
 - We propose a new framework for radiation oncology
 - Genomic-based
 - Non-uniform benefit
 - Biologically optimized, individualized RT dose
 - Integrates interaction with immune system
 - Opens the door to precision genomic radiation therapy



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