A Biomarker that can predict which indication will be approved by the FDA

by Steen Knudsen, Anker Hansen, Thomas Jensen, Jon Askaa and Peter Buhl Jensen

Medical Prognosis Institute, Venlighedsvej 1, 2970 Hørsholm, Denmark.



1. Cancer cell lines show differential sensitivity to cancer drugs	2. Sensitivity pattern is related to drug mode of action	3. Explanatory model based on gene expression		
Belinostat (HDAC inhibitor) Afinitor (mTOR inhibitor)	SAHA	OMICS		
GI ₅₀ Mean Graph for Compound 726630 NCI Cancer Screen Current Data Average GI ₅₀ over all cell lines is 1.78E-7 M <u>Cell Panel</u> <u>Cell Line</u> <u>Log GI₅₀ GI₅₀ <u>CCRF-CEM</u> -7.7 <u>HL-60(TB)</u> -7.4 <u>K-562</u> -7.2 <u>MOLT-4</u> -7.6 <u>RPMI-8226</u> -8.0 <u>RPMI-8226</u> -8.0 <u>RPMI-826</u> -8.0 <u>RPMI-8226</u> -8.0 <u>RPMI-8226</u> -8.0 <u>RPMI-826</u> -8.0</u>	Y - O erlotinib oxali belinostat entinostat iressa	Clinical		
Non-Small Cell LungSR 4549/ATCC-7.8 -6.4 HOP-62Non-Small Cell LungSR 4549/ATCC-8.0 EKVXnumber of drugs for their ability to inhibit their ability to inhibitNon-Small Cell LungNon-Small Cell LungNCI-H226-6.2NCI-H232-6.5NCI-H232Non-Small Cell LungNon-Small Cell LungNon-Small Cell LungNCI-H232-7.1NCI-H322-6.5NCI-H322-6.5NCI-H322Non-Small Cell LungNon-Small Cell LungNCI-H232-7.1NCI-H460-6.8NCI-H322-6.8NCI-H322Non-Small Cell LungNon-Small Cell Lung<	Cisplatin	In silico systems biology		
HCC-2998 HCT-116 HCT-116 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-15 HCT-16 HCT-15 HCT-16 HCT-15 HCT-15 HCT-16 HCT-15 HCT-1	gelda carboplatin tomoirolimuo			







Drug Response Predictor (DRP)

If a growth inhibition vector (as shown in red in panel 1) is combined with baseline transcriptomics and systems biology, a model results that can predict a patient's clinical response to the drug based on a pre-treatment biopsy.

4. Summary of 24 clinical validation trials: 20 positive

Cancer F	Patients	Drug(s)	Patent	Primary	P value ORR	ORR	ORR Bottom 10%	ORR Top 10%	Meta-analysis of
				(sec) endpoint					
Breast	268	tamoxifen	Issued	RFS	0.03*				24 clinical trials
Breast	136	tamoxifen	Issued	DMFS	0.03*				ing Fisher's com
Breast	102	16 combinations	Issued	DMFS	0.006*				
DLBCL	166	СНОР	Issued	CR (OS)	0.007*				bined probability
DLBCL	414	(R)-CHOP	Issued	OS	1e-15*				vields a n-value o
Breast	244	11 combinations	Issued	pCR	8e-12*	20%	8%	32%	
Breast	125	TET/FEC	Issued	pCR	0.007*	44%	31%	62%	1e-35.
Breast	24	docetaxel	Issued	pCR	0.02*	46%	0%	100%	
DLBCL (miRNA)	116	R-CHOP/CHOEP	Issued	CR	0.03*	89%	82%	100%	
Hodgkin	130	ABVD	Issued	CR	0.003*	71%	62%	92%	Stoumer's Z-scor
AML	13	Belinostat+idarub.	Issued	ORR	0.02*	40%	0%	100%	weighed by numl
AML	88	7 combinations	Issued	CR	0.02*				of notionts in one
Breast	44	Oncology Drug X	Pending	CR	0.01*	55%	0%	100%	of patients in eac
NSCLC	21	Tarceva (erlotinib)	Pending	PFS	0.02*				study gives a p-v
NSCLC	50	cisplatin	Issued	OS	0.03*				$\int \frac{1}{29} \frac{2}{36}$
Breast	24	cisplatin	Issued	Miller-Payne	0.02*	63%	67%	100%	0120-30.
Ovarian	28	cisplatin	Issued	OS	0.06				
Breast	114	epirubicin	Pending	pCR (DMFS)	0.9 (0.03)	14%	8%	25%	
AML	53	decitabine	Issued	ORR	0.01*	50%	25%	75%	
Breast	19	Anastrozole	Pending	ORR	0.9				
AML	79	HAM	Issued	CR	0.45				
Myeloma	84	VAD	Issued	CR	0.004*	95%	~50%	100%	
ALL	161	Methotrexate	Issued	WBC count	0.008*				
Myeloma	169	bortezomib	Issued	ORR	0.008*	50%	23%	29%	

5. Explanatory model can predict cancer type with highest response rate



This study was repeated with 3 other drugs: decitabine, vorinostat and sorafenib and similar correlations were observed in all three cases.

6. Explanatory model can predict which indication will be approved by the FDA

7. Future perspectives



Rank of the FDA approved indication for 65 cancer drugs. For each drug, a biomarker was developed and used to predict the relative sensitivity of 27 cancer indications to the drug. The figure shows the rank of the indication approved by the FDA for all 65 drugs. Where more than one indication has been approved, the highest rank is shown. In vitro data for nine drugs (14%) failed quality criteria (red circles). Among the remaining, the approved indication was correctly predicted for 12 drugs (21%, P=0.0005). The approved indication was among the top 5 predictions for 32 drugs (57%, P=0.002). A random prediction would on average rank at 10.9 (orange line). 73% of the predictions (green circles) were better than random. While there is an overweight of drugs approved for leukemia and lymphoma on the left, and an overweight of drugs approved only for solid tumors on the right, simply predicting hematological cancers to be most sensitive would fare no better than random prediction (orange line).

MPI has developed biomarkers for six anticancer drugs currently in the pipeline at biopharmaceutical companies, allowing for a future prospective test of their ability to predict an indication that will later be approved by the FDA (although a development bias cannot be ruled out).

8. References

Rank of the FDA approved indication

1. Lars Bullinger, Steen Knudsen, Herve Dombret et al. Results of a phase I/II trial of belinostat in combination with idarubicin in AML – favorable impact on mainly intermediate cytogenetic risk AML can be predicted by gene expression profiling. Presented at ESMO 2012, Vienna, Austria

2. Wang W, Baggerly KA, Knudsen S, Askaa J, Mazin W, Coombes KR. Independent validation of a model using cell line chemosensitivity to predict response to therapy. J Natl Cancer Inst. 2013 Sep 4;105(17):1284-91

3. Pieter Sonneveld, Mark Vanduin, Steen Knudsen, et al. A novel chemosensitivity index for chemotherapy correlates to myeloma cell lines resistance and patients refractoriness. Presented at XXII International Myeloma Workshop in Washington, DC, Feb 26, 2009

4. Junjie Chen, Steen Knudsen, Wiktor Mazin, Jesper Dahlgaard, and Baolin Zhang. A 71-gene signature of TRAIL sensitivity in cancer cells. (October 25, 2011); Mol Cancer Ther, 10.1158/1535-7163.

1EDICAL PROGNOSIS INSTITUTE

