Abstract #1254

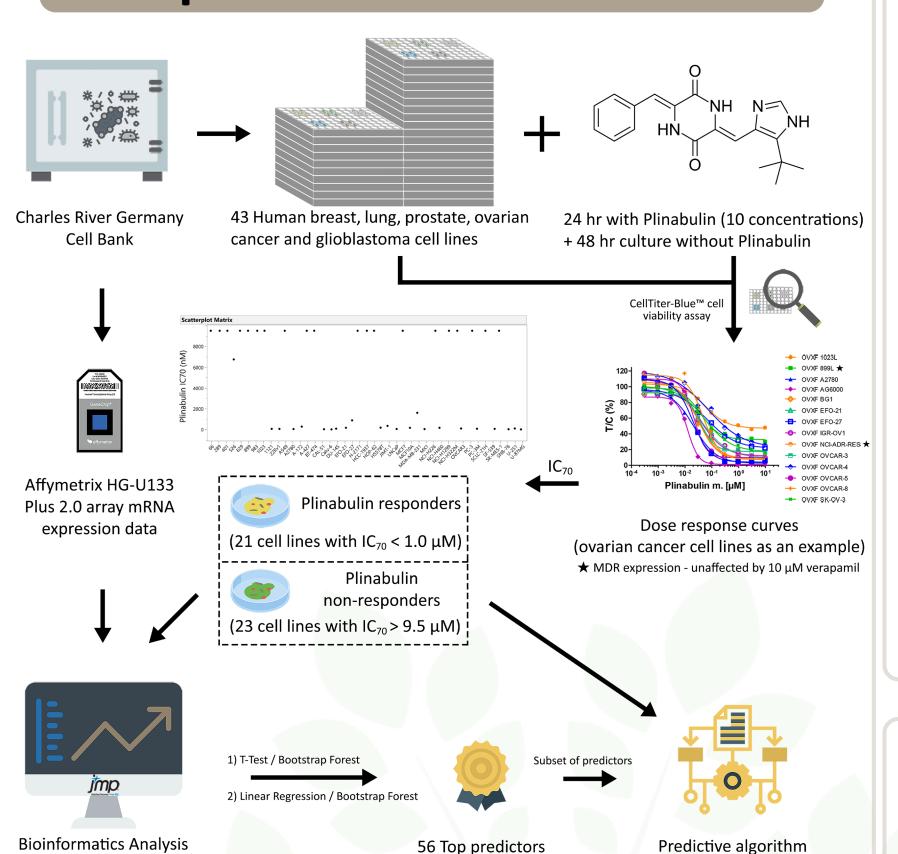
Predictive Models for Tumor Cell Targeting with Plinabulin, Derived from *In Vitro* Screening and Affymetrix mRNA Expression Data



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Experimental flow chart



Aim and methods

A i see

Tubulin binding drugs are approved for the treatment of many cancer types, without the use of molecular markers to select patients likely to respond. Plinabulin binds β-tubulin at a differentiated site and is being tested in a Phase 3 clinical study for the treatment of NSCLC. Additional cancer indications are being considered for plinabulin and an algorithm for selecting especially sensitive cancers and patient subgroups would be of significant value.

Summary of Methods

Affymetrix HG-U133 Plus 2.0 array mRNA expression data for 43 human breast, lung, prostate, ovarian or CNS cancer cell lines was utilized to develop mathematical models to predict *in vitro* plinabulin potency against the same cell lines. To characterize potency, cells were treated for only 24 hours with plinabulin and then cultured for another 48 hours without plinabulin. Viable cell number was then measured with a Cell Titer-Blue Assay, and the plinabulin concentration causing a 70% reduction in signal (IC₇₀) versus vehicle treated controls was derived. Cell lines were clearly separable into plinabulin Active (21 cell lines with IC₇₀ < 1.0 μ M) and Inactive (IC₇₀ > 9.5 μ M) groups. Log2 transformed Affymetrix gene probeset signal values, preprocessed with the GeneChip robust multi-array average analysis algorithm, were then used to select the top predictor probesets, utilizing JMP 14.1 statistical software.

TTest-Predictor Method: The probesets ranked among the top 200 predictors with a bootstrap forest partitioning technique were compared by t-test in Active versus Inactive tumor cell lines. For those reaching p < 0.05, the annotated genes for these probesets, if available, were noted. Next, all of the probesets in the array that are mapped to the same noted genes were identified. Jetset scoring methods to assess each probeset for specificity, splice isoform coverage, and robustness against transcript degradation have been shown to be valuable tools in assessing the value of each probeset, in particular correlating with protein expression (Li et al., 2011). At this point therefore, the probeset with the highest Jetset score that mapped to each noted gene, with a p value < 0.01 for Active versus Inactive values, was selected. In addition, probesets without a mapped gene, with a p value < 0.01 for plinabulin Active versus Inactive values, were also selected. 40 top probesets (HITs) were selected in this manner.

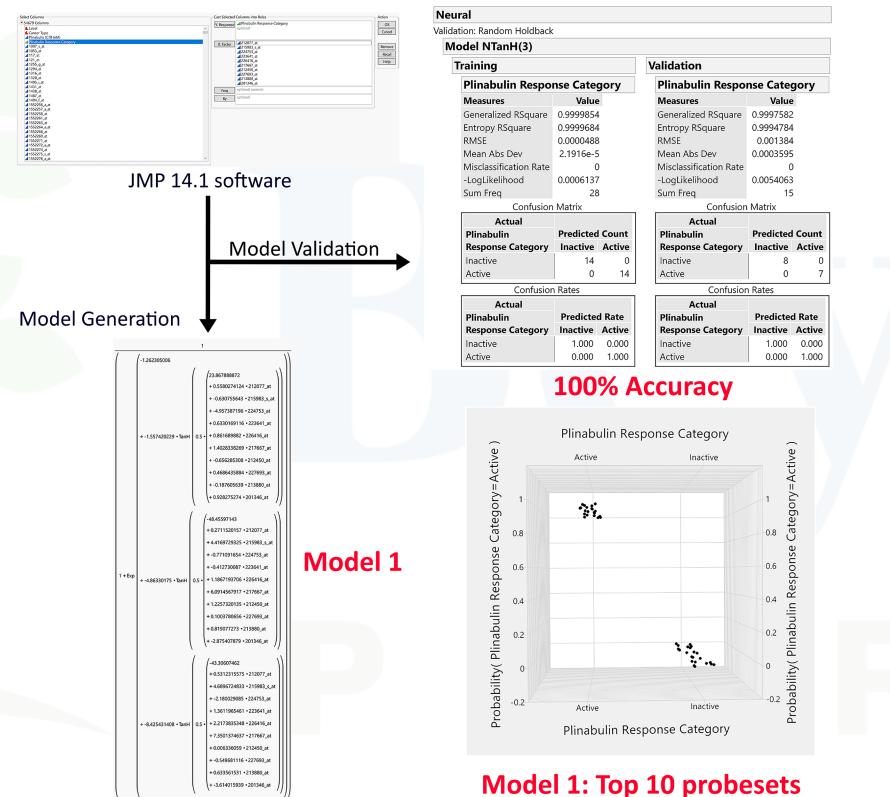
Correlation-Predictor Method: Probesets with correlation p < 0.01 versus plinabulin IC₇₀ were ranked for their ability to predict plinabulin Active versus Inactive 3 times. For those with an average rank < 50 (low score = high rank) and not already picked up with the TTest-Predictor method above, the gene annotation was evaluated. 16 "HIT" probesets that were non-annotated or had the highest Jetset score for each identified gene, and had differential expression between plinabulin Active versus Inactive cell lines (p < 0.01), were selected.

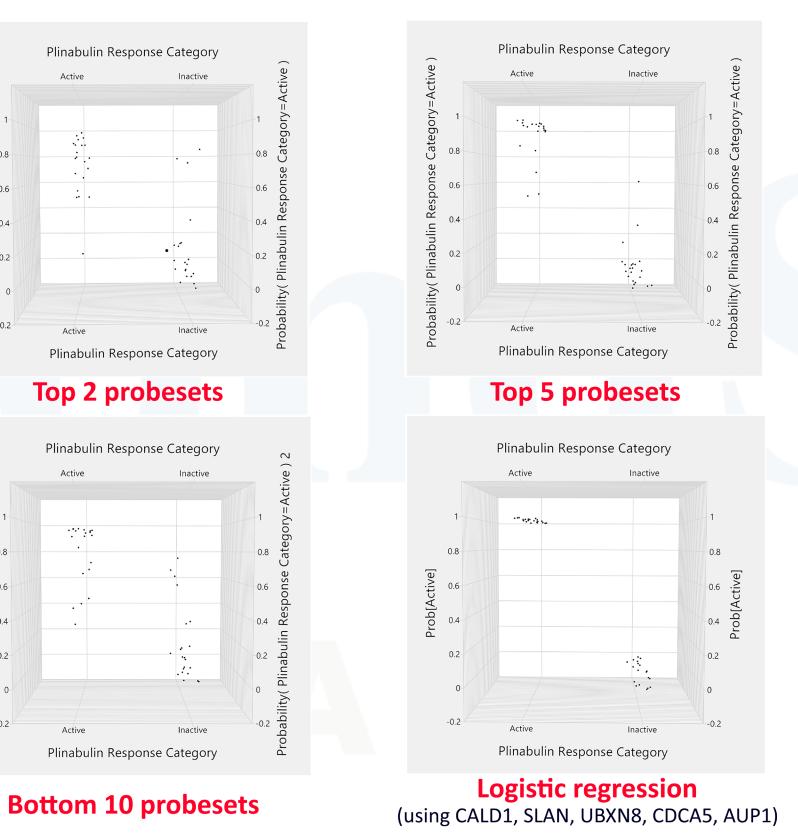
The 56 HITs were ranked 4 times as predictors. Models (mathematical algorithms) were constructed from the 56 HIT gene probesets in JMP to identify plinabulin responding cell lines, utilizing either one-layer TanH multimode fit neural networks or binary logistic regression.

Top predictors for Plinabulin response

Avg. Rank	Gene Symbol	\triangle_{Active}	Biology/Function	Avg. Rank	Gene Symbol	\triangle_{Active}	Bio
1.00	CALD1	1	Caldesmon; calmodulin/actin binding; Regulate cell morphology and motility	28.00	Unannotated	1	
2.50	UBXN8 (REP8/UBXD6)	个	p97 ATPase complex cofactor; Lipid droplet biogenesis; Misfolded proteins repair	29.75	AUP1	↓ ↓	Lipid
2.50	CDCA5	\downarrow	Sororin; Sister chromatid cohesion, separation	30.50	ZFX	1	Tran
5.75	Unannotated	个		34.75	MRPL30	1	Mito
6.50	ERI1	\uparrow	RNA exonuclease; Histone mRNA degradation; rRNA processing	35.25	TRAK1	1	Mito
7.25	SEC14L1P1	\uparrow	SEC14 like 1 pseudogene 1	35.50	RCCD1	↓	Hist
7.25	SECISBP2L (SLAN)	\uparrow	SECIS Binding Protein 2; Centrosome maturation; Retard proliferation, inhibit Aurora A	36.50	Unannotated	1	
7.50	WDR20	\downarrow	USP12-UAF1 deubiquitinating enzyme complex; Tumor suppressor inhibiting ERK & AKT	37.25	ZMAT3 (WIG1)	1	TP5
8.50	LGR5	\uparrow	GPCR with ligand R-spondin; Intestinal stem cell activation; Wnt signaling	37.75	GEMIN7	↓	Part
10.75	ADIPOR2	\downarrow	Adiponectin receptor protein 2; Insulin resistance or hepatic fat accumulation	38.00	ZNF106 (ZFP106)	1	Sens
11.00	RUFY2	\uparrow	RUN and FYVE domain containing 2; Detected in brain, lung, testis	39.00	Unannotated	1	
11.25	COL5A2	\uparrow	Fibrillar forming collagen; High expression in bladder cancer, breast cancer, CRC	39.25	GLT8D1	1	Gly
12.25	YTHDC2	\uparrow	Oncogene induced by c-Jun, ATF2; Regulate mRNA transcripts	42.25	CASC4	1	Spli
14.75	RPL12	\downarrow	Bind to 26S ribosomal RNA; Regulate translation interacting with eEF1A and eEF2	42.50	FAM98B	1	tRN
18.50	MTMR9	\uparrow	Myotubularin-related protein; Obesity, hypertension; PtdIns(3)P regulation	42.75	NME1-NME2	↓	Nuc
18.50	Unannotated	个		44.00	ноокз	1	Mic
18.50	Unannotated	个		44.50	CSTF3 (CSTF77)	1	Poly
18.75	TM9SF3	个	Bind to β-adrenergic ligands; Tumor invasion; Transporter in spermatogenesis	44.50	ACTR3	1	Con
21.00	CALB2	\downarrow	Calretinin; calcium-binding; Ca ²⁺ homeostasis; Apoptosis; Cell adhesion, cell growth	45.50	Unannotated	1	
21.25	Unannotated	个		46.00	RPL38	↓	60S
21.25	WDR92	\downarrow	Protein with 2 WD40 repeat domains; Modulator of apoptosis and dynein assembly	46.75	Unannotated	1	
22.75	DGUOK	\downarrow	Deoxyguanosine Kinase; Hypercerebral mitochondrial DNA depletion syndrome	46.75	PLOD1	1	Coll
23.00	CTNNB1	\uparrow	β-catenin; Wnt signaling; Cell adhesion and gene transcription; Oncogene	48.25	MRAS	1	GTF
23.25	FKBP4	\downarrow	Immunophilin protein family; Enhance progestin receptor-mediated transcription	49.25	ZNF441	1	Trar
23.25	BRPF3	\downarrow	Histone H3 acetyltransferase activity; DNA replication and H3K14 acetylation (S phase)	51.00	RELB	↓	NF-I
23.50	DENND2D	\downarrow	GEF; Regulate Rab9a/b GTPases and lysosome distribution	54.75	NLE1	↓	Not
26.00	TMEM47	\uparrow	Regulate epithelial cell junction; Associate actomyosin structures, cell morphology	54.75	Unannotated	↓	
27.25	RPS19	\downarrow	Pre-rRNA processing (40S subunits)	55.50	MRPS23	↓ ↓	Mite

Predictive model development





id droplet clustering → VLDL↑; Degradation of misfold ER proteins

ct of pre-mRNA splicing complex (survival of motor neuron complex

nponent of ARP2/3 complex; Formation of the mitotic actin cluster

Pase in TNF- α , MAPK signaling; Regulate RAF kinase activation

nscriptional regulation; Downregulated in ERG high tumors

ch activity; Axial skeletal formation in mice model

S ribosomal protein; Regulate axial skeletal patterning (Hox protein) in mice

llagen fibrils cross-linking; Specify ECM facilitating organs & tissues morphogenesis

tochondrial ribosomal protein (28S subunit); Cancer cell proliferation and metastasis

ochondrial ribosomal protein (39S subunit); Mitochondrial translation

stone demethylase activity; Regulate microtubule stability; Cell cycle repression

53-dependent pathway; Cell cycle arrest; MYC-N activation; Tumor suppressor

nsory and motor neuron maintenance; mitochondrial electron transport chain

licing; Upregulated in HER-2/Neu overexpressed cell line; Breast & Ovarian Cancer NA-splicing ligase complex; Correlated with PRMT1 (arginine methyltransferase) cleoside diphosphate kinase; DNA cleavage; DNA-binding transcription factors crotubule-dependent intracellular vesicle & protein trafficking; Centrosome assembles.

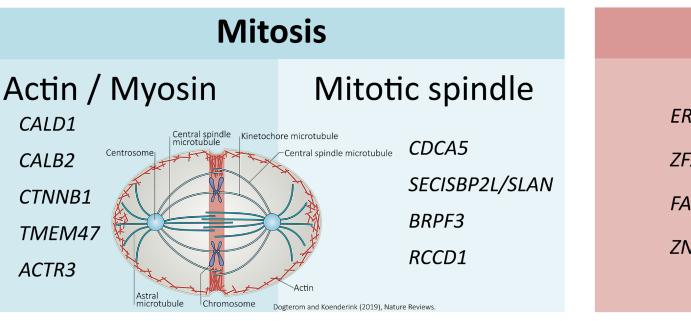
nscriptional coactivator; Cell proliferation; Apoptosis

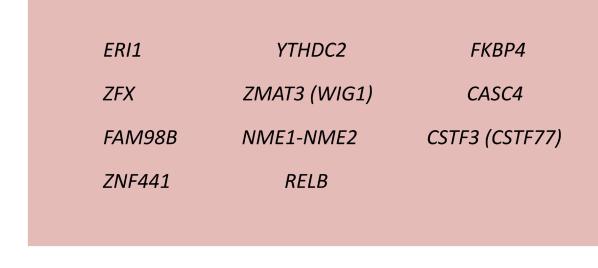
cosyltransferase family; Schizophrenia risk gene

Plinabulin activity for different cancer cell lines

Cancer Type	Plinabulin Activity	43 Cell Lines Tested for Predictive Model Development	Additional Cell Lines Tested	
CNS	Active	A-172, SF-539, SNB-78, U-251, U-87MG	LN-229, SF-268	
CNS	Inactive		M059K ($IC_{70} > 10 \mu M$, $IC_{50} = 87 nM$)	
Lung	Active	Calu-6, LXFL1121, NCI-H460		\
Lung	Inactive	LXFA289, LXFA526, LXFA629, LXFA983, A549, HOP-62, NCI-H322M,		Model 1
		NCI-H226, SK-MES-1, A427, NCI-H1299, H2171, SCLC-21H, LXFE66NL		predicte
Breast	Active	CAL-51, HS578T, JIMT-1, MCF10A, MX1		all 3 to b
Breast	Inactive	MAXFTN401, BT-474, HCC-1937, MCF7, MDA-MB-231		active
Ovarian	Active	A2780, EFO-21, EFO-27, OVCAR3		
Ovarian	Inactive	OVXF899, OVXF1023		
Prostate	Active	22Rv1, DU-145, LNCaP, PC-3M		
Prostate	Inactive	PC-3		

Predictors of Plinabulin activity grouped by biological/cellular function





mRNA regulation

Ribosomal function									
40S subunit	60S subunit	Mitochondrial							
RPS10	RPL12 #	MRPL30 #							
RPS19 #	RPL38 # RPL13P5	MRPS23 #							
RACK1	RPL18 RPL8								
→ All 400/C00		on in Dinabulia							

responding cell lines

Among 56 top predictors

Conclusion

Predictive models:

- Models have been developed that predict the ability of plinabulin to target cancer cells *in vitro* with high confidence utilizing mRNA expression data **Future steps:**
- Test algorithms for the ability to predict plinabulin activity in vitro with additional cancer cell lines, as well as plinabulin in vivo activity
- Utilize top predictors to better understand plinabulin's MOA
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