Evaluation of Action Mechanisms of Toxic Chemicals Using JFCR39, a Panel of Human Cancer Cell Lines[S]

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ABSTRACT

We previously established a panel of human cancer cell lines, JFCR39, coupled to an anticancer drug activity database; this panel is comparable with the NCI60 panel developed by the National Cancer Institute. The JFCR39 system can be used to predict the molecular targets or evaluate the action mechanisms of the test compounds by comparing their cell growth inhibition profiles (i.e., fingerprints) with those of the standard anticancer drugs using the COMPARE program. In this study, we used this drug activity database-coupled JFCR39 system to evaluate the action mechanisms of various chemical compounds, including toxic chemicals, agricultural chemicals, drugs, and synthetic intermediates. Fingerprints of 130 chemicals were determined and stored in the database. Sixty-nine of 130 chemicals ($\sim$60%) satisfied our criteria for the further analysis and were classified by cluster analysis of the fingerprints of these chemicals and several standard anticancer drugs into the following three clusters: 1) anticancer drugs, 2) chemicals that shared similar action mechanisms (for example, ouabain and digoxin), and 3) chemicals whose action mechanisms were unknown. These results suggested that chemicals belonging to a cluster (i.e., a cluster of toxic chemicals, a cluster of anticancer drugs, etc.) shared similar action mechanism. In summary, the JFCR39 system can classify chemicals based on their fingerprints, even when their action mechanisms are unknown, and it is highly probable that the chemicals within a cluster share common action mechanisms.

Determining the action mechanism or identifying the molecular target of a chemical with pharmacological activity or adverse side effects is highly desirable. Although various test methods are currently available for determining the action mechanisms of chemicals, such as methods based on animal models, methods based on cellular models, bacterial mutagenicity test, the uterotropic assay (Kanno et al., 2002), Hershberger test (Hershberger et al., 1953), and the reporter assay for the nuclear receptor agonists, determination of the action mechanisms of pharmacologically active chemicals, including the toxic chemicals, is still a difficult and challenging task. Therefore, it is highly desirable to develop efficient test methods for evaluating toxicity of chemicals.

A number of screening methods are currently available for discovering new anticancer drugs. One very powerful and unique approach using multiple cancer cell lines was developed at NCI (Paull et al., 1989; Weinstein et al., 1992, 1997) and in our laboratory (Yamori et al., 1999; Dan et al., 2002, 2003; Yamori, 2003; Nakatsu et al., 2005; Akashi and Yamori, 2007; Akashi et al., 2007; Nakamura et al., 2007). This bioinformatics-based approach enables mechanism-oriented evaluation of anticancer drugs. For example, we can evaluate the cell toxicity in vitro by determining the 50% growth inhibition (GI50), total growth inhibition, and 50% lethal concentration across a panel of 39 human cancer cell lines (JFCR39). We can also predict the molecular targets or evaluate the action mechanisms of the test compounds by comparing the cell growth inhibition profiles (termed "fingerprints") across the panel for these compounds with those of

ABBREVIATIONS: GI50, 50% growth inhibition concentration; GI50, 50% growth inhibition; SN-38, 7-ethyl-10-hydroxycamptothecin; SV-NN, snake venom from *N. nigricollis*; SV-NNK; snake venom from *N. naja kaouthia*.

TABLE 1
List of chemicals tested. Chemical names, abbreviations, and applications/targets/mechanisms of the test compounds are summarized.

JCI No	Name	Abbreviation	Application/Target/Mechanism
−691	Trioctyltin	TOT	Organotin
−690	Triphenyltin	TPT	Organotin
−689	Dibutyltin		Organotin
−688	AM-580		RARα
−687	TTNPB		RAR
−686	13-*cis* Retinoic acid	13-*cis*	RAR
−607	Methoprene		Agricultural chemical
−606	Methoprene acid		RXR
−605	5-aza-2′-deoxycytidine	5-AzaC	Methylation
−604	Carbaryl		Agricultural chemical
−603	Acephate		Agricultural chemical
−602	Sodium arsenite		Agricultural chemical
−601	Testosterone propionate	TP	Testosterone
−600	Ethynyl estradiol	EE	Estrogenic
−599	Thiram		Agricultural chemical
−598	Dimethylformamide	DMF	Solvent
−568	α-Bungarotoxin	αBuTX	Neurotoxin
−567	Snake venom from *Trimeresurus flavoviridis*	SV-TF	Snake venom
−566	Snake venom from *Crotalus atrox*	SV-CA	Snake venom
−565	Snake venom from *Agkistrodon halys blomhoffii*	SV-AHB	Snake venom
−564	Dexamethasone	DEX	Steroid
−563	3-Methylcholanthrene	3-MC	Teratogenicity/carcinogenicity
−562	N-Ethyl-N-nitrosourea	ENU	Teratogenicity/carcinogenicity
−561	Diethylnitrosamine	DEN	Teratogenicity/carcinogenicity
−560	All *trans*-retinoic acid	ATRA	RAR + RXR
−559	9-cis Retinoic acid	9-cis	RAR
−558	Levothyroxine	T4	Thyroid hormone
−557	3-Amino-1*H*-1,2,4-triazole	3AST	Agricultural chemical
−555	2-Vinylpyridine	2VP	Synthetic intermediate
−553	Phenobarbital	PB	Antiepileptic
−552	Acetaminophen	APAP	Analgesic
−551	Isoniazid		Phthisic
−549	4-Ethylnitrobenzene	4ENB	Synthetic intermediate
−548	1,2-Dichloro-3-nitrobenzene	1,2DC3NB	Pigment/synthetic intermediate
−546	N-Methylaniline	NMA	Synthetic intermediate
−545	2-Aminomethylpyridine	2AMP	Synthetic intermediate
−544	1*H*-1,2,4-Triazole		Synthetic intermediate
−543	1*H*-1,2,3-Triazole		Synthetic intermediate
−542	4-Amino-2,6-dichlorophenol	4A2,6DCP	Synthetic intermediate
−541	2,4-Dinitrophenol	2,4 DNP	Agricultural chemical
−513	Capsaicin		Food constituent
−485	2-Methoxyestradiol		Estrogenic
−466	Colcemid		Spindle inhibitor
−465	2,4-Dinitrochlorobenzene	2,4DCB	Pigment/mutagenesis
−464	Troglitazone		Diabetic
−463	Clofibrate		Antilipemic
−459	Bis(2-ethylhexyl)phthalate	DEHP	Plasticizer
−458	Thiourea		Agricultural chemical
−447	Cacodylic acid		Agricultural chemical
−446	Amitrole		Agricultural chemical
−445	4-Octylphenol	OP	Reproductive effector
−444	2,6-Dimethylaniline	2,6-Xylidene	Natural product
−443	1,2-Dibromo-3-chloropropane	DBCP	Agricultural chemical
−442	1,1-Dimethylhydrazine	1,1DMH	Reproductive effector
−441	Sulfanylamide		Agricultural chemical
−440	Streptozotocin		Agricultural chemical
−439	Spironolactone		Aldosterone antagonist
−438	*para*-Aminoazobenzene	pAAB	Pigment/mutagenicity/carcinogenicity
−437	*para*-Cresidine		Pigment/carcinogenicity
−436	Neostigmine bromide		Parasympathomimetics
−435	*para*-Dichlorobenzene	pDCB	Pigment/Agricultural chemical
−434	Phenytoin		Antiepileptic
−433	*ortho*-Toluidine	oToluidine	Pigment
−432	Imipramine		Antidepressant
−431	Cobalt chloride		Teratogenicity/mutagenicity
−428	Atrazine		Agricultural chemical
−427	Propylthiouracil		Teratogenicity/carcinogenicity
−426	Thalidomide (L + D)		Teratogenicity
−425	Carbon tetrachloride	CCl$_4$	Teratogenicity/carcinogenicity
−424	Hydroquinone		Oxidative stress
−423	Monocrotaline		Mutagenicity/carcinogenicity
−422	Vinyl chloride		Carcinogenicity
−421	Tributyltin chloride	TBT	Ship bottom paint/organotin
−420	Valproic acid		Antiepileptic
−419	Benzene		Carcinogenicity

TABLE 1—(*Continued*)

JCI No	Name	Abbreviation	Application/Target/Mechanism
−418	Acrylamide		Neurotoxin/carcinogenicity
−417	Hexachlorobenzene	BHC	Agricultural chemical/carcinogenicity
−346	2-Deoxyglucose	2-DG	Glycolytic pathway/glycosylation inhibitor
−325	Pentachlorophenol	PCP	Agricultural chemical/teratogenicity/carcinogenicity
−324	Aniline		Oxidative stress/methemoglobinemia/carcinogenicity
−323	Triazine		Agricultural chemical
−322	Edifenphos	EDDP	Agricultural chemical/antibiotics/choline esterase
−321	γ-1,2,3,4,5,6-Hexachlorocyclohexane	γ-BHC	Agricultural chemical/carcinogenicity
−320	Dichlorvos	DDVP	Agricultural chemical/teratogenicity/carcinogenicity
−319	*O*-Ethyl *O*-4-nitrophenyl phenylphosphonothioate	EPN	Agricultural chemical
−318	Cadmium chloride	CdCl$_2$	Teratogenicity/carcinogenicity
−317	Phenylmercury acetate	PMA	Fungicides/mutagenicity
−316	Mercaptoacetic acid		Synthetic intermediate
−315	1,3-Diphenylguanidine	DPG	Vulcanizing agent
−314	3,4,4′-Trichlorocarbanilide	TCC	Cosmetics/antibacterial agent
−313	3-Iodo-2-propynyl butylcarbamate	IPBC	Antibacterial agent
−311	2,3,3,3−2′,3′,3′,3′-Octachlorodipropylether	S-421	Agricultural chemical/antibacterial agent
−310	1,2-Benzisothiazolin-3-one	BIT	Antibacterial agent
−309	Isobornylthiocyanoacetate	IBTA	Antibacterial agent
−308	*p*-Chlorophenyl-3-iodopropargylformal	CPIP	Antibacterial agent
−307	Zinc butylxanthate	ZBX	Vulcanizing agent
−306	Polypropylene glycol	PG	Synthetic intermediate
−305	10,10′-Oxy-bis(phenoxyarsine)	OBPA	Antibacterial agent
−296	Snake venom from *Naja naja kaouthia*	SV-NNK	Snake venom
−295	Snake venom from *Naja nigricollis*	SV-NN	Snake venom
−294	2,5-di(*tert*-butyl)-1,4-Hydroquinone	DTBHQ	Oxidative stress
−293	Ibotenic Acid		Mushroom toxin/neurotoxin
−292	*N*-Methy-4-phenyl-1,2,3,6-tetrahydropyridine	MPTP	Neurotoxin
−289	Tetrodotoxin		Natural product/Na$^+$ channel inhibitor
−288	ICI 182,780		Estrogen antagonist
−275	Benzophenone		Agricultural chemical
−274	1,2-Dibromo-3-chloropropane	DBCP	Antibacterial agent/insecticide/carcinogenicity
−273	Zineb		Agricultural chemical
−272	Dieldrin		Insecticide
−271	Hexachlorobenzene	HCB	Antibacterial agent/carcinogenicity
−270	Ziram		Antibacterial agent/vulcanizing agent
−269	Chlordane		Insecticide/carcinogenicity
−268	4,4′-Dichlorodiphenyltrichloroethane	p,p′-DDT	Insecticide/carcinogenicity/teratogenicity
−267	Bisphenol A	BPA	Estrogenic
−266	17-β-Estradiol	E2	Estrogenic
−265	Diethylstilbestrol	DES	Estrogenic
−261	Paraquat		Agricultural chemical/oxidative stress
−247	Ouabain		Cardiac glycosides
−245	Okadaic acid		Natural product/PP1, PP2A inhibitor
−242	Antimycin A1		Agricultural chemical
−232	Digoxin		Cardiac glycosides
−201	OH-Flutamide		Flutamide derivative/androgen antagonist
−200	Flutamide		Anticancer drugs/androgen antagonist
−185	30% H$_2$O$_2$		Oxidative stress
−182	*N*-Acetyl-L-cysteine	NAC	Super oxyside scavenger
−181	L-Ascorbic acid		Food constituent
−179	Dopamine		Neurotransmitter
−177	Caffeine		Food constituent
−168	Cycloheximide		Protein synthesis inhibitor
−144	4-Hydroxyphenylretinamide	4-HPR	RAR
−137	Indomethacin		COX inhibitor
−99	SN-38		Irinotecan derivative/Topo I
−96	Toremifene		Anticancer drugs/estrogen antagonist
−95	Tamoxifen		Anticancer drugs/estrogen antagonist
−63	Cyclosporin A		Anticancer drugs/helper T cell
−46	HCFU		Anticancer drugs/antimetabolite(pyrimidine)
−36	Docetaxel		Anticancer drugs/tubulin
−35	Paclitaxel		Anticancer drugs/tubulin
−34	Colchicine		Antipodagric/tubulin
−33	Cisplatin		Anticancer drugs/DNA cross linker
−32	Carboplatin		Anticancer drugs/DNA cross linker
−31	Irinotecan		Anticancer drugs/Topo I
−30	Camptothecin	CPT	Anticancer drugs/Topo I
−24	Methotrexate		Anticancer drugs/DHFR
−19	Vincristine		Anticancer drugs/tubulin
−18	Vinblastine		Anticancer drugs/tubulin
−16	Mitomycin-C	MMC	Anticancer drugs/DNA alkylator
−9	Tegafur		Anticancer drugs/antimetabolite(pyrimidine)
−8	5-Fluorouracil	5-FU	Anticancer drugs/antimetabolite(pyrimidine)
−5	Cytarabine		Anticancer drugs/antimetabolite(pyrimidine)
−4	Nitrogen mustard		Anticancer drugs/DNA alkylator

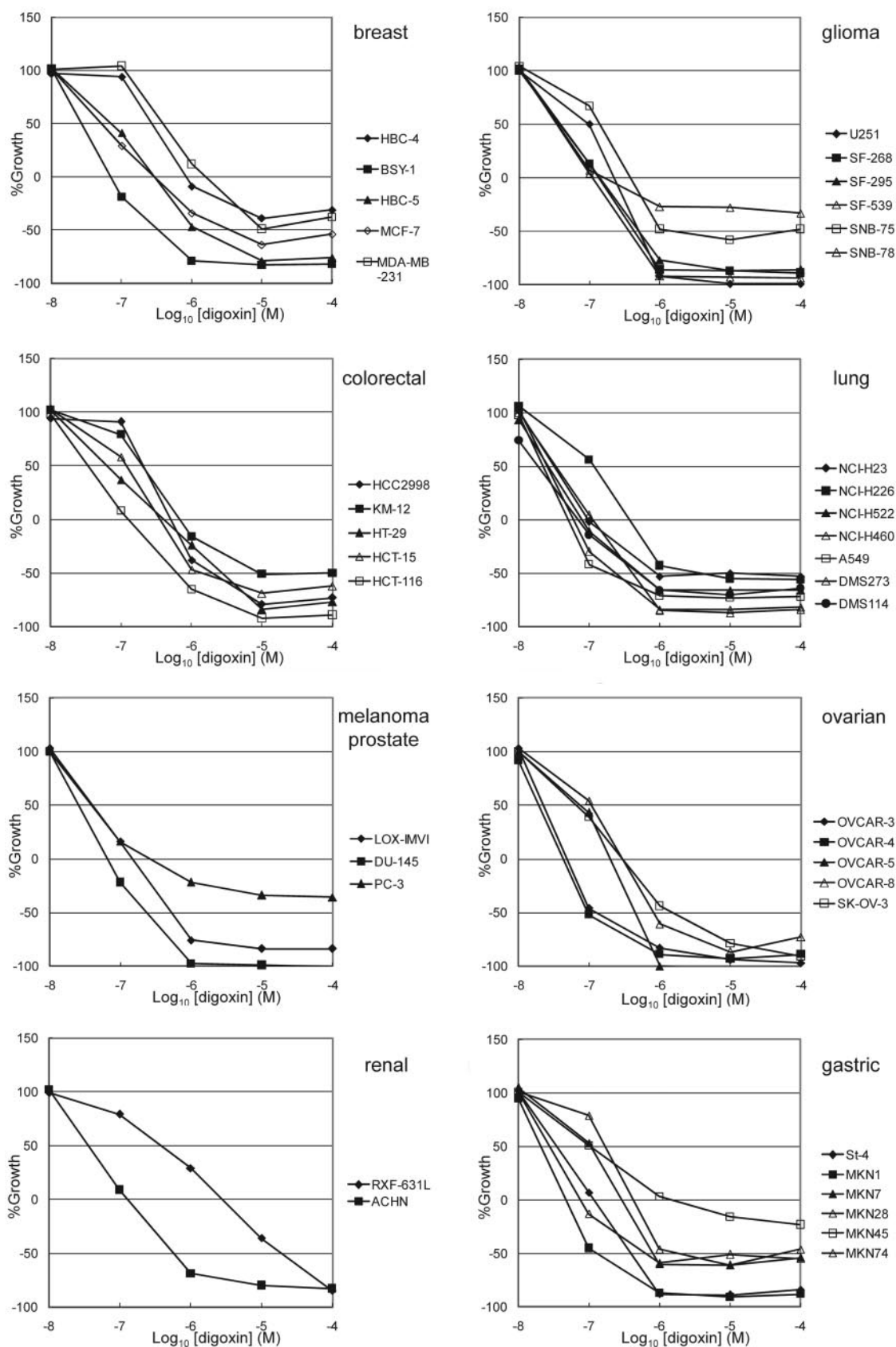RAR, retinoic acid receptor; RXR; retinoid X receptor.

Fig. 1. Dose response curves of digoxin against growth of JFCR-39 cells. The *x*-axis represents concentration of digoxin and the *y-axis* represents percentage growth. The GI50 represents the concentration required to inhibit cell growth by 50% compared with untreated controls.

the standard anticancer drugs using the COMPARE algorithm (Yamori et al., 1999). We have used this system successfully and demonstrated that the molecular targets of the novel chemicals MS-274, FJ5002, and ZSTK474 were topoisomerases I and II (Yamori et al., 1999), telomerase (Naasani et al., 1999), and phosphatidylinositol 3-kinase (Yaguchi et al., 2006), respectively. Several other interesting studies, based on a panel of cancer cells, classified anticancer drugs according to their action mechanisms or molecular targets by cluster analysis of their GI50 values (Weinstein et al., 1992, 1997; Dan et al., 2002). Correlation analysis has also been used to explore the genes associated with the sensitivity of the cells in the panel to anticancer drugs (Scherf et al., 2000; Okutsu et al., 2002; Zembutsu et al., 2002; Nakatsu et al., 2005).

In this study, we have examined the potential of the JFCR39 system in classifying various chemicals, and predicted their action mechanisms. For this purpose, we have determined the fingerprints of 130 different types of chemicals including toxic chemicals, pesticides, drugs and synthetic intermediates, and then classified these chemicals according to the cluster analysis of their fingerprints.

Materials and Methods

Cell Lines and Cell Cultures. The panel of human cancer cell lines has been described previously (Yamori et al., 1999; Dan et al., 2002) and consists of the following 39 human cancer cell lines: lung cancer, NCI-H23, NCI-H226, NCI-H522, NCI-H460, A549, DMS273, and DMS114; colorectal cancer, HCC-2998, KM-12, HT-29, HCT-15, and HCT-116; gastric cancer, MKN-1, MKN-7, MKN-28, MKN-45, MKN-74, and St-4; ovarian cancer, OVCAR-3, OVCAR-4, OVCAR-5, OVCAR-8, and SK-OV-3; breast cancer, BSY-1, HBC-4, HBC-5, MDA-MB-231, and MCF-7; renal cancer, RXF-631L and ACHN; melanoma, LOX-IMVI; glioma, U251, SF-295, SF-539, SF-268, SNB-75, and SNB-78; and prostate cancer, DU-145 and PC-3. All cell lines were cultured in RPMI 1640 medium (Nissui Pharmaceutical, Tokyo, Japan) with 5% fetal bovine serum, penicillin (100 units/ml), and streptomycin (100 μg/ml) at 37°C under 5% CO_2.

Determination of Cell Growth Inhibition Profiles. Growth inhibition experiments were performed to assess the sensitivity of the cells to various chemicals as described before (Yamori et al., 1999; Dan et al., 2002). Growth inhibition was measured by determining the changes in the amounts of total cellular protein after 48 h of chemical treatment using a sulforhodamine B assay. For each chemical, the growth assay was performed using a total of five different concentrations of the chemical (for example, 10^{-4}, 10^{-5},
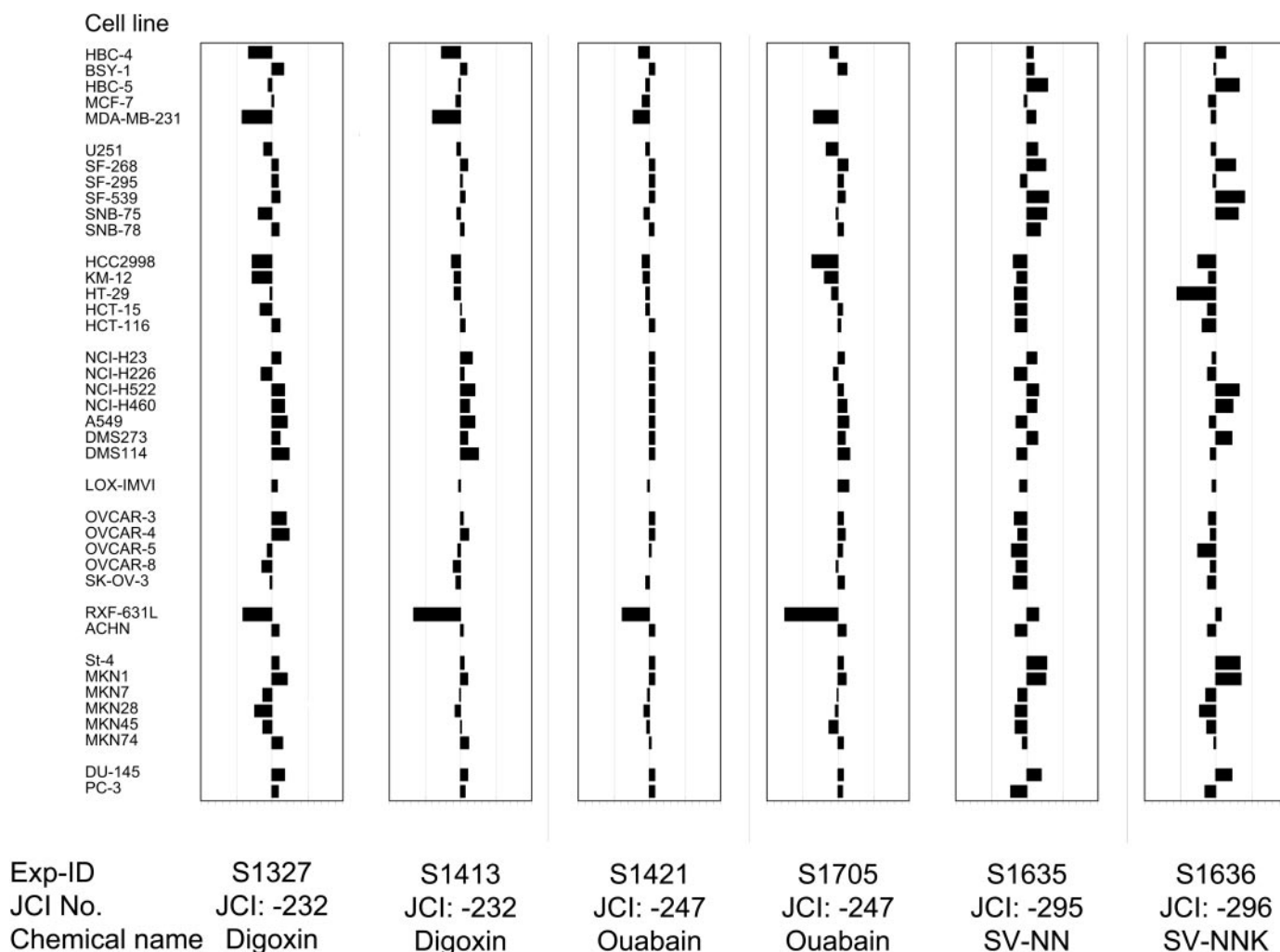
Fig. 2. Fingerprints of digoxin, ouabain, SV-NN, and SV-NNK. Fingerprint shows the differential growth inhibition pattern of the cells in the JFCR-39 panel against the test chemical. The X-axis represents relative value of GI50; $(-1) \times (\log \text{GI50} - \text{MG-MID})$; MG-MID is the mean value of the log GI50. Zero means the mean GI50 and one means the GI50 value is 10-fold more sensitive than the mean GI50. Exp-ID and JCI numbers are the ID for the experiment and ID for the chemical, respectively, in our database.

10^{-6}, 10^{-7}, and 10^{-8} M) and one negative control. All assays were performed in duplicate. This GI50 calculation method is well established and reliable through anticancer drug screen using NCI60 as well as JFCR39 (Paull et al., 1989; Yamori et al., 1999; Yamori, 2003). At each test concentration, the percentage growth was calculated using the following seven absorbance measurements: growth at time 0 (T0), growth of the control cells (C), and test growth in the presence of five different concentrations (T) of a drug. The percentage growth inhibition was calculated as: % growth = $100 \times [(T - T0)/(C - T0)]$ when $T \geq T0$, and % growth = $100 \times [(T - T0)/T]$ when $T < T0$. The GI50 values, which represent 50% growth inhibition concentration, were calculated as $100 \times [(T - T0)/(C - T0)] = 50$. When the GI50 of a chemical could not be calculated, the highest used concentration was assigned as its GI50 value. Absolute values of GI50 were then log transformed for further analysis. We certified the accuracy of measured GI50 data by using reference control chemicals, such as mitomycin-C, paclitaxel, and SN-38, in every experiment and by checking the dose response curves.

Chemicals. Spironolactone, *para*-aminoazobenzene, *para*-cresidine, neostigmine bromide, *para*-dichlorobenzene, phenytoin, *ortho*-toluidine, imipramine, cobalt chloride, atrazine, propylthiouracil, (D,L)-thalidomide, carbon tetrachloride, hydroquinone, monocrotaline, vinyl chloride, tributyl-tin chloride, valproic acid, benzene, acrylamide, pentachlorophenol, aniline, 1,3-diphenylguanidine, polypropylene glycol, 10,10′-oxy-bis(phenoxyarsine), testosterone propionate, carbaryl, acephate, bisphenol A, 17-β-estradiol, diethylstilbestrol, and α-bungarotoxin were purchased from Wako (Tokyo, Japan). Snake venoms from *Agkistrodon halys blomhoffii*, *Trimeresurus flavoviridis*, *Crotalus atrox*, *Naja nigricollis*, and *Naja naja kaouthia* were purchased from Latoxan (Valence, France).

2-Aminomethylpyridine, 1*H*-1,2,4-triazole, 1*H*-1,2,3-triazole, 3,4,4′-trichlorocarbanilide, edifenphos, dichlorvos, *O*-ethyl *O*-4-nitrophenyl phenylphosphonothioate, 2,4-dinitrophenol, *N*-methylaniline, 1,2-dichloro-3-nitrobenzene, 4-ethylnitrobenzene, 2-vinylpyridine, 3-amino-1*H*-1,2,4-triazole, *N*-ethyl-*N*-nitrosourea, 5-aza-2′- deoxycytidine, ethynyl estradiol, 3-methylcholanthrene, phenobarbital, acetaminophen, isoniazid, capsaicin, *N*-deacetyl-*N*-methylcolchicine (Colcemid), 2.4-dinitrochlorobenzene, and dexamethasone were from Sigma Chemicals (St. Louis, MO). Methoprene acid, methoprene, all-*trans* retinoic acid, and 9-cis retinoic acid were from BIOMOL International L.P. (Plymouth Meeting, PA). Levothyroxine was from MP Biomedicals (Irvine, CA). 3-Iodo-2-propynyl butylcarbamate was from Olin Japan Inc. (Tokyo, Japan), *p*-chlorophenyl-3-iodopropargylformal was from Nagase ChemteX (Osaka, Japan), and 2,3,3,3–2′,3′,3′,3′-octachlorodipropylether was from Sankyo Chemical Industries, Ltd. (Tokyo, Japan). 1,2-Benzisothiazolin-3-one was from Riverson (Osaka, Japan), zinc butylxanthate was from Ouchishinko Chemical Industrial Co., Ltd. (Tokyo, Japan), and 4-amino-2,6-dichlorophenol was from Tokyo Kasei Kogyo Co. Ltd. (Tokyo, Japan).

Hierarchical Clustering. Hierarchical clustering analysis was carried out using the average linkage method and the "GeneSpring" software (Silicon Genetics, Inc., Redwood, CA). Pearson correlation coefficients were used to determine the degree of similarity.

Results

Sensitivity of JFCR39 to Chemicals. Sensitivity of the JFCR39 panel of cells to 130 chemicals was determined as described under *Materials and Methods*. Table 1 summarizes

TABLE 2

Log$_{10}$ GI50 values of chemicals for each cell line in the JFCR-39 panel

Hi-conc means the highest concentration of the test chemical used. When the growth inhibition was over 50% at the Hi-Conc, GI$_{50}$ was assigned the Hi-Conc value.

Exp-ID	S3416	S3415	S3413	S3245	S3117	S3414	S3118	S3246	S3125	S3124	S3123	S1636	S1635	S1634	S1718
JCI No	−687	−686	−559	−559	−559	−560	−560	−560	−567	−566	−565	−296	−295	−294	−294
Name or Abbr.	TTNPB	13-*cis*	9-*cis*			ATRA			SV-TF	SV-CA	SV-AHB	SV-NNK	SV-NN	DTBHQ	
Hi-Conc.	−4	−4	−4	−4	−4	−4	−4	−4	−4	−4	−4	−4	−4	−4	−4
HBC-4	−4.76	−4.00	−4.53	−4.40	−4.43	−4.42	−4.41	−4.41	−5.87	−5.80	−5.66	−7.25	−7.31	−4.72	−4.80
BSY-1	−4.78	−4.16	−4.60	−4.73	−4.73	−4.69	−4.70	−4.81	−6.31	−6.06	−5.76	−6.93	−7.34	−5.07	−4.93
HBC-5	−4.80	−4.41	−4.56	−4.57	−4.61	−4.61	−4.47	−4.51	−6.98	−6.45	−5.73	−7.64	−7.72	−4.89	−4.78
MCF-7	−4.73	−4.35	−4.40	−4.39	−4.48	−4.48	−4.54	−4.66	−5.87	−5.78	−5.68	−6.77	−7.08	−5.29	−5.25
MDA-MB-231	−4.75	−4.21	−4.70	−4.55	−4.69	−4.63	−4.53	−4.65	−5.90	−5.86	−5.84	−6.84	−7.39	−5.52	−5.30
U251	−4.77	−4.14	−4.61	−4.51	−4.61	−4.57	−4.45	−4.63	−6.45	−5.76	−5.70	−6.85	−7.44	−4.96	−5.11
SF-268	−4.75	−4.00	−4.24	−4.55	−4.40	−4.47	−4.48	−4.76	−5.90	−5.79	−5.70	−7.53	−7.67	−4.77	−4.81
SF-295	−4.80	−4.29	−4.54	−4.66	−4.60	−4.59	−4.48	−4.57	−6.19	−5.80	−5.74	−6.89	−6.97	−4.87	−4.97
SF-539	−4.95	−4.35	−4.75	−4.80	−4.79	−4.80	−4.71	−4.76	−6.39	−5.96	−5.81	−7.79	−7.75	−4.79	−4.86
SNB-75	−5.31	−5.28	−5.13	−5.19		−4.71	−4.69	−4.87	−6.41	−6.33	−5.93	−7.60	−7.70	−4.67	−4.80
SNB-78	−4.77	−4.25	−4.69	−4.78	−4.86	−4.49	−4.70	−4.68	−6.19	−6.00	−5.95	−6.97	−7.53	−4.75	−4.75
HCC2998	−4.68	−4.00	−4.48	−4.61	−4.62	−4.55	−4.62	−4.76	−5.91	−5.75	−5.67	−6.47	−6.77	−4.82	−4.75
KM-12	−4.70	−4.00	−4.46	−4.51	−4.48	−4.51	−4.47	−4.58	−5.93	−5.80	−5.65	−6.77	−6.87	−4.74	−4.77
HT-29	−4.73	−4.00	−4.47	−4.53	−4.50	−4.60	−4.52	−4.56	−5.90	−5.80	−5.56	−5.89	−6.78	−4.80	−4.89
HCT-15	−4.72	−4.25	−4.45	−4.49	−4.48	−4.52	−4.57	−4.53	−5.88	−5.76	−5.57	−6.73	−6.82	−4.72	−4.77
HCT-116	−4.77	−4.07	−4.67	−4.59	−4.67	−4.71	−4.61	−4.64	−6.46	−6.10	−5.77	−6.58	−6.82	−4.98	−5.13
NCI-H23	−4.74	−4.00	−4.47	−4.60	−4.59	−4.61	−4.55	−4.63	−6.11	−5.75	−5.72	−6.86	−7.42	−4.76	−4.90
NCI-H226	−4.72	−4.00	−4.61	−4.68	−4.78	−4.80	−4.54	−5.48	−5.95	−5.81	−5.76	−6.73	−6.78	−4.89	−4.91
NCI-H522	−4.72	−4.45	−4.68	−4.82	−4.77	−4.71	−4.71	−4.68	−6.45	−5.99	−5.78	−7.62	−7.46	−5.37	−5.37
NCI-H460	−4.70	−4.00	−4.55	−4.63	−4.58	−4.68	−4.55	−4.49	−5.96	−5.82	−5.72	−7.44	−7.42	−4.84	−4.84
A549	−4.79	−4.00	−4.72	−4.77	−4.78	−4.70	−4.62	−4.53	−5.91	−5.79	−5.71	−6.80	−6.83	−4.83	−4.87
DMS273	−4.57	−4.21	−4.50	−4.62	−4.55	−4.57	−4.51	−4.49	−6.20	−5.81	−5.72	−7.43	−7.44	−4.91	−4.98
DMS114	−4.77	−4.16	−4.33	−4.62	−4.49	−4.51	−4.53	−4.61	−6.66	−6.33	−5.77	−6.83	−6.88	−5.12	−5.21
LOX-IMVI	−4.77	−4.69	−4.68	−4.66	−4.70	−4.77	−4.74	−4.74	−6.75	−6.59	−5.76	−6.86	−6.94	−5.05	−5.15
OVCAR-3	−4.77	−4.38	−4.56	−4.67	−4.72	−4.64	−4.62	−4.71	−6.61	−6.13	−5.89	−6.77	−6.79	−4.89	−4.86
OVCAR-4	−4.72	−4.05	−4.63	−4.64	−4.64	−4.58	−4.39	−4.54	−6.73	−6.23	−5.80	−6.82	−6.90	−5.13	−4.90
OVCAR-5	−4.75	−4.00	−4.33	−4.39	−4.42	−4.44	−4.34	−4.44	−5.92	−5.74	−5.67	−6.46	−6.71	−5.22	−5.26
OVCAR-8	−4.75	−4.23	−4.50	−4.53	−4.59	−4.66	−4.67	−4.70	−5.95	−5.77	−5.69	−6.82	−6.84	−4.64	−4.70
SK-OV-3	−4.79	−4.00	−4.49	−4.51	−4.81	−4.52	−4.54	−4.50	−5.76	−5.64	−4.91	−6.75	−6.76	−4.64	−4.74
RXF-631L	−4.77	−4.00	−4.54	−4.58	−4.60	−4.72	−4.63	−4.61	−5.91	−5.80	−5.59	−7.13	−7.46	−4.81	−4.84
ACHN	−4.73	−4.00	−4.56	−4.66	−4.56	−4.50	−4.40	−4.76	−5.90	−5.79	−5.73	−6.74	−6.80	−4.71	−4.83
St-4	−4.74	−4.00	−4.42	−4.54	−4.65	−4.53	−4.49	−4.57	−5.91	−5.81	−5.76	−7.65	−7.70	−4.68	−4.75
MKN1	−4.75	−4.33	−4.56	−4.63	−4.62	−4.56	−4.45	−4.48	−6.15	−5.81	−5.78	−7.67	−7.68	−4.59	−4.81
MKN7	−4.78	−4.40	−4.68	−4.59	−4.70	−4.73	−4.56	−4.65	−6.29	−5.85	−5.76	−6.70	−6.90	−4.79	−4.84
MKN28	−4.71	−4.28	−4.56	−4.59	−4.59	−4.65	−4.56	−4.60	−6.10	−5.93	−5.68	−6.51	−6.81	−4.72	−4.89
MKN45	−4.72	−4.00	−4.51	−4.41	−4.46	−4.73	−4.41	−4.43	−6.06	−5.90	−5.69	−6.71	−6.82	−4.71	−4.87
MKN74	−4.74	−4.40	−4.61	−4.63	−4.61	−4.73	−4.68	−4.67	−5.97	−5.92	−5.61	−6.92	−7.00	−5.10	−5.42
DU-145	−4.68	−4.00	−4.25	−4.78	−4.41	−4.42	−4.44	−4.54	−6.08	−5.82	−5.75	−7.43	−7.55	−4.59	−5.02
PC-3	−4.74	−4.00	−4.58	−4.65	−4.48	−4.74	−4.39	−4.51	−5.83	−5.77	−5.61	−6.67	−6.69	−4.89	−4.74

abbreviations, applications, targets, and known mechanisms of 130 chemicals and 21 anticancer drugs. Approximately 15% of the chemicals were assessed twice or more. Approximately 40% of the chemicals tested had little effect on the growth of cells in the JFCR39 panel. However, the rest of the chemicals significantly inhibited the cell growth across the JFCR39 panel. For example, Fig. 1 shows the dose response curves of the cells in the JFCR39 panel against digoxin. The concentration at which the cell growth is inhibited by 50% represents GI50. Figure 2 shows the fingerprints of four chemicals [digoxin, ouabain, snake venom from *N. nigricollis* (SV-NN), and snake venom from *N. naja kaouthia* (SV-NNK)], which differentially inhibited the growth of cells in the JFCR39 panel; these fingerprints were drawn based on a calculation using a set of GI50s and clearly represented the GI50 pattern. These results were highly reproducible in that the Pearson correlation coefficient of the duplicate experiments for digoxin was 0.839 ($p < 0.001$) and that for ouabain was 0.864 ($p < 0.001$). It is noteworthy that, digoxin and ouabain, both of which are cardiac glycosides and inhibit Na-K ATPase, showed similar fingerprints. The fingerprints of SV-NNK and SV-NN, which belong to the elapidae, known as cobras, were also similar, but were different from the fingerprints of digoxin and ouabain. Table 2 summarizes only a portion of the GI50 values from 160 experiments involving 130 chemicals and 42 experiments involving 21 anticancer drugs. GI50 values from all experiments are described in the

Supplemental Data (Table S1). All these data were stored in a chemosensitivity database and used for further analysis.

Classification of the Chemicals by Hierarchical Clustering. Sixty-nine chemicals were selected for further analysis based on the following criteria: 1) GI50 values for the test chemical can be determined for at least 10 cell lines in the JFCR39 panel, and 2) the range of log GI50 for the test chemical is more than 0.6, suggesting differential growth inhibition. We analyzed the GI50 values of these 69 chemicals and 20 anticancer drugs by hierarchical clustering analysis (Fig. 3). We found approximately 12 clusters (threshold: $r = 0$, Fig. 3, clusters A–L), which were further divided into 49 subclusters (threshold: $r = 0.408$, Fig. 3, clusters A1–L6).

Analysis of Clusters. Most anticancer drugs we have tested belonged either to cluster A or cluster H, depending on their modes of action (Dan et al., 2002). The targets of the anticancer drugs belonging to the cluster A were related to DNA (Topo I, antimetabolite of pyridine, DNA alkylator) and the target of the anticancer drugs belonging to the cluster H was tubulin. We presently found that cisplatin exceptionally belonged to cluster F2, not cluster A, although it is known to cross-link DNA strands (Jamieson and Lippard, 1999; Wong and Giandomenico, 1999). We were also able to precisely group the clusters into several subclusters having similar characteristics. For example, the cardiac glycosides digoxin and ouabain were grouped in one cluster (cluster F3). SV-

| S3243	S3244	S1534	S3237	S3238	S1525	S3236	S1928	S1421	S1705	S1327	S1413	S1413	S3408	S3409	S2421
−599	−599	−270	−270	−270	−261	−261	−261	−247	−247	−232	−232	−232	−421	−421	−421
Thiram		Ziram			Paraquat			Ouabain		Digoxin			TBT		
−4	−4	−4	−4	−4	−4	−3	−4	−4	−6	−4	−4	−4	−4	−4	−4
−4.71	−4.79	−5.80	−5.73	−5.70	−4.00	−3.61	−4.00	−7.54	−7.28	−6.57	−6.96	−6.96	−6.79	−6.77	−6.72
−6.97	−7.12	−6.85	−6.76	−6.60	−4.00	−4.45	−4.51	−8.00	−7.76	−7.58	−7.68	−7.68	−7.03	−7.01	−6.83
−7.41	−7.66	−7.18	−7.47	−7.47	−4.68	−4.70		−7.76	−7.51	−7.15	−7.44	−7.44	−6.76	−6.88	−6.83
−4.77	−4.80	−6.00	−5.84	−5.83	−4.06	−3.72	−4.00	−7.64	−7.51	−7.29	−7.39	−7.39	−6.86	−6.84	−6.79
−4.66	−4.68	−5.64	−5.75	−5.63	−4.00	−3.57	−4.00	−7.40	−6.81	−6.41	−6.72	−6.72	−6.83	−6.81	−6.70
−4.75	−4.78	−5.71	−5.79	−5.82	−4.00	−3.69	−4.00	−7.75	−7.16	−7.01	−7.40	−7.40	−6.79	−6.77	−6.72
−4.86	−4.96	−5.74	−5.83	−7.01	−4.00	−4.08	−4.00	−8.00	−7.77	−7.42	−7.70	−7.70	−6.84	−6.85	−6.71
−4.77	−4.89	−5.71	−5.70	−5.79	−4.47	−4.37	−4.20	−8.00	−7.64	−7.42	−7.55	−7.55	−6.75	−6.73	−6.76
−4.75	−4.88	−5.73	−5.75	−5.77	−4.00	−4.03	−4.00	−8.00	−7.70	−7.46	−7.63	−7.63	−6.77	−6.72	−6.67
−4.71	−4.96	−5.79	−5.92	−5.80	−4.00	−3.94	−4.00	−7.70	−7.45	−6.86	−7.40	−7.40	−6.99	−6.95	−7.05
−4.70	−4.78	−5.69	−5.64	−5.69	−4.00	−3.78	−4.00	−7.98	−7.64	−7.45	−7.60	−7.60	−6.72	−6.79	−6.70
−4.82	−4.69	−5.76	−5.79	−5.81	−4.00	−3.70	−4.00	−7.64	−6.77	−6.68	−7.25	−7.25	−6.77	−6.79	−6.72
−4.80	−4.80	−5.43	−5.74	−5.73	−4.00	−3.58	−4.00	−7.67	−7.12	−6.69	−7.34	−7.34	−7.00	−6.98	−6.74
−4.68	−4.85	−5.75	−5.77	−5.76	−4.10	−4.03	−4.07	−7.75	−7.31	−7.20	−7.34	−7.34	−6.89	−6.84	−6.66
−4.68	−4.75	−5.70	−5.72	−5.83	−4.00	−3.64	−4.00	−7.74	−7.63	−6.92	−7.54	−7.54	−6.88	−6.84	−6.70
−4.72	−4.72	−5.74	−5.68	−5.77	−4.00	−3.60	−4.00	−8.00	−7.57	−7.47	−7.62	−7.62	−6.90	−6.85	−6.74
−4.69	−4.78	−5.96	−5.85	−5.84	−4.19	−4.18	−4.00	−8.00	−7.67	−7.50	−7.84	−7.84	−6.90	−6.85	−6.76
−6.33	−6.74	−5.63	−5.96	−6.12	−4.41	−4.41	−4.00	−8.00	−7.37	−6.93	−7.61	−7.61	−6.99	−6.91	−6.74
−7.49	−7.50	−7.44	−7.66	−8.00	−4.49	−4.71	−4.59	−8.00	−7.64	−7.59	−7.91	−7.91	−6.83	−6.80	−6.25
−6.14	−6.16	−6.30	−6.10	−6.15	−4.30	−4.45	−4.37	−8.00	−7.74	−7.60	−7.77	−7.77	−6.98	−6.98	−6.56
−4.84	−4.82	−5.97	−5.91	−5.91	−4.49	−4.49	−4.41	−8.00	−7.80	−7.66	−7.91	−7.91	−6.82	−6.87	−6.73
−6.64	−6.58	−6.43	−6.84	−6.82	−4.25	−4.43	−4.30	−8.00	−7.71	−7.48	−7.72	−7.72	−6.74	−6.75	−6.70
−7.18	−7.39	−7.37	−7.38	−7.43	−4.50	−4.63	−4.27	−8.00	−7.84	−7.73	−8.00	−8.00	−7.11	−7.12	−7.02
−4.68	−4.71	−5.66	−5.71	−5.70	−4.00	−3.51	−4.00	−7.80	−7.80	−7.39	−7.46	−7.46	−6.93	−6.94	−6.76
−4.86	−6.25	−6.07	−6.35	−6.32	−4.00	−4.46	−4.28	−8.00	−7.66	−7.64	−7.59	−7.59	−6.80	−6.82	−6.74
−4.77	−6.67	−5.90	−5.91	−5.87	−4.00	−4.21	−4.48	−8.00	−7.71	−7.71	−7.72	−7.72	−6.91	−7.20	−6.80
−4.90	−6.00	−6.11	−6.91	−6.74	−4.00	−3.98	−4.00	−7.89	−7.62	−7.12	−7.42	−7.42	−6.90	−6.93	−6.75
−4.62	−4.74	−5.59	−5.68	−5.67	−4.00	−3.84	−4.00	−7.85	−7.44	−6.97	−7.29	−7.29	−6.73	−6.67	−6.57
−4.39	−4.38	−4.92	−5.49	−5.53	−4.00	−3.39	−4.00	−7.74	−7.67	−7.18	−7.38	−7.38	−6.80	−6.77	−6.68
−4.75	−4.65	−5.57	−5.63	−5.60	−4.00	−3.60	−4.00	−7.10	−6.00	−6.42	−6.20	−6.20	−6.78	−6.76	−6.68
−4.52	−4.60	−5.64	−5.68	−5.69	−4.00	−3.51	−4.00	−8.00	−7.72	−7.44	−7.59	−7.59	−6.78	−6.77	−6.76
−4.59	−4.72	−5.99	−5.73	−5.81	−4.00	−3.58	−4.00	−8.00	−7.65	−7.45	−7.60	−7.60	−6.80	−6.80	−6.72
−4.80	−4.85	−6.82	−5.84	−5.92	−4.41	−4.61	−4.48	−8.00	−7.72	−7.68	−7.72	−7.72	−7.25	−7.15	−6.87
−4.79	−4.82	−6.56	−5.84	−5.82	−4.08	−4.29	−4.32	−7.80	−7.48	−6.98	−7.47	−7.47	−7.34	−7.07	−6.86
−7.18	−7.21	−5.82	−7.09	−7.13	−4.00	−4.40	−4.18	−7.70	−7.42	−6.77	−7.37	−7.37	−6.84	−6.82	−6.87
−6.65	−6.71	−6.05	−7.18	−6.86	−4.00	−4.29	−4.35	−7.77	−7.25	−6.99	−7.52	−7.52	−6.96	−6.97	−6.87
−7.05	−7.08	−6.35	−5.86	−7.05	−4.00	−4.47	−4.06	−7.91	−7.65	−7.55	−7.74	−7.74	−6.97	−7.37	−7.05
−4.47	−4.70	−5.68	−5.68	−5.65	−4.00	−3.57	−4.00	−8.00	−7.64	−7.59	−7.71	−7.71	−6.90	−6.89	−6.70
−4.42	−4.77	−5.61	−5.53	−5.56	−4.00	−3.64	−4.37	−8.00	−7.62	−7.41	−7.62	−7.62	−6.77	−6.78	−6.73
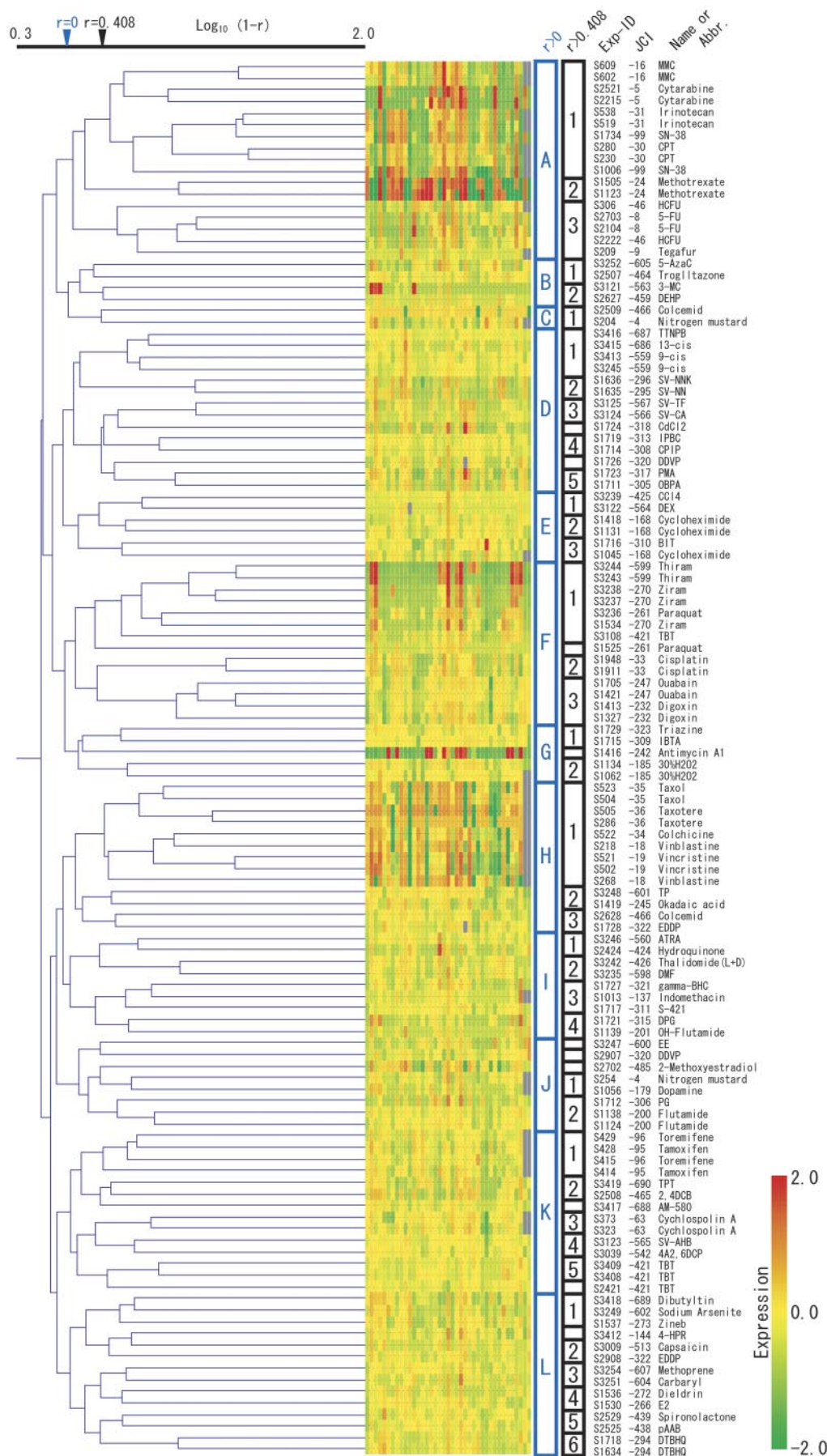
Fig. 3. Hierarchical clustering of 69 test chemicals and 20 anticancer drugs based on their GI50 values. Hierarchical clustering method was an "average linkage method" using the Pearson correlation as distance. We classified the chemicals into two kinds of clusters; their threshold values were $r = 0$ and $r = 0.408$ ($p < 0.01$), respectively. Gradient color indicates relative level (log transformed) of GI50. Red, more sensitive than the mean GI50 (2.0); yellow, mean GI50 (0.0); and green, less sensitive than the mean GI50 (-2.0). On the color scale, red represents the GI50 value that is 100-fold higher than the mean GI50.

NNK and SV-NN, on the other hand, belonged to the cluster D2. These results are in accordance with the similar fingerprints shown in Fig. 2. It is noteworthy that the snake venoms from the *C. atrox* and *T. flavoviridis*, species belonging to the viperidae family of snakes, formed another cluster (cluster D3), which was different from that of the elapidae family of snakes, *N. naja kaouthia* and *N. nigricollis*. 9-*cis* Retinoic acid, 13-*cis* retinoic acid, and 4-[*E*-2-(5,6,7,8-tetrahydro-5,5,8,8-tetra-methyl-2-naphthalenyl)-1-propenyl]benzoic acid, which are RAR agonists (Aström et al., 1990), also formed a separate cluster (cluster D1). Likewise, agricultural chemicals paraquat, ziram, and thiram formed a single cluster (cluster F1).

Discussion

The JFCR39 system coupled to a drug activity database is a good model for investigating the diversity of chemosensitivity in cancer cells. We have previously established panels of human cancer cell lines [JFCR39 (Yamori, 2003) and JFCR45 (Nakatsu et al., 2005)]. We used these panels of cells to demonstrate that they provide powerful means to predict the action mechanisms of drugs, and also used them to identify new target compounds. In this manuscript, we used the JFCR39 system to evaluate various chemicals (such as toxic chemicals, agricultural chemicals, and synthetic intermediates), which are not anticancer drugs, and classified them according to their molecular target or action mechanism. As a result, these chemicals were classified into a number of clusters. Our results also suggested that each cluster consisted of chemicals sharing a common action mechanism.

We determined the growth inhibition of cells in the JFCR39 panel by 130 chemicals and calculated their GI50 values. Some of the chemicals were assessed twice or more to confirm the reproducibility of the assay. We had to exclude 61 chemicals from further analysis because they did not inhibit the cells in the JFCR39 panel significantly. The rest of the chemicals (69 of 130, ~60%) met our selection criteria and were evaluated by cluster analysis.

First, we found that the chemicals tested in duplicate formed tight clusters, showing high reproducibility. Next, we investigated the difference between these 69 test chemicals and the anticancer drugs. Sixty-nine chemicals, which are not anticancer drugs, formed several clusters, which were different from the anticancer drug clusters. These results suggest that the action mechanisms of these chemicals are different from the action mechanisms of the anticancer drugs. However, we found that cisplatin did not belong to the cluster A, which consisted of DNA-targeting anticancer drugs. We do not understand the reason at present. However, there is a possibility that cisplatin has other action mechanisms, which may have made the fingerprint of cisplatin different from those of other DNA-targeting drugs. Indeed, cisplatin is known to form DNA-protein cross-links (Zwelling et al., 1979; Chválová et al., 2007).

Our analysis also identified several interesting clusters. For example, the cluster F3 consisted of cardiac glycosides digoxin and ouabain, both of which inhibit Na-K ATPase (Reuter et al., 2002). The cluster D1 consisted of 9-*cis* retinoic acid, 13-*cis* retinoic acid, and 4-[*E*-2-(5,6,7,8-tetrahydro-5,5,8,8-tetra-methyl-2-naphthalenyl)-1-propenyl]benzoic acid, which are RAR agonists. These results suggest that chemicals other

than the anticancer drugs also form clusters when they share the same action mechanisms. It is noteworthy that SV-NNK and SV-NN, from snakes that belonged to the elapidae family, formed one cluster (cluster D2). In contrast, the snake venoms from the *C. atrox* and *T. flavoviridis*, which belonged to the viperidae family, formed a cluster (cluster D3) different from the elapidae cluster. These results are reasonable because snake venoms from different snake families are known to differ not only in composition but also in levels of toxicity and mechanisms of action.

The agricultural chemicals paraquat, ziram, and thiram were also classified into a single cluster (cluster F1). Among these agricultural chemicals, the action mechanism of ziram is not known. However, both paraquat and thiram are known to induce oxidative stress (Cereser et al., 2001; Suntres, 2002). Therefore, based on our observations, we could suggest that ziram also acted by inducing oxidative stress. The agricultural chemicals methoprene (insect growth regulator) and carbaryl (chorine esterase inhibitor) formed cluster L3, although their common mechanism is unknown. Cluster D4 and D5 consist of the antibacterial agents or fungicides. 3-Iodo-2-propynyl-butylcarbamate and *p*-chlorophenyl-3′-iodopropargylformal, belonging to cluster D4, are the iodo-type antibacterial agents.

Thus, cluster analysis of GI50 values of various chemicals, determined using the JFCR39 cell panel, suggests that the JFCR39 system could, at least in part, allow classification of chemical compounds on the basis of their action mechanisms. Our analysis also suggests that the chemicals belonging the same cluster share a common action mechanism. We are going to develop a larger library of reference chemicals with known action mechanisms (i.e., various inhibitors of biological pathways), and expand our database by integrating their GI50 measurements, which will make the cluster analysis as well as the COMPARE analysis more informative for predicting the mechanism of test chemicals.

In conclusion, to evaluate the potential of the JFCR39 system in predicting the action mechanisms of toxic chemicals, we investigated the fingerprints of 130 different types of chemical compounds including toxic chemicals, pesticides, drugs, and synthetic intermediates. Using the hierarchical clustering analysis, we classified 69 chemicals, at least in part, based on their action mechanisms. Thus, this approach using the JFCR39 cell panel is useful not only in predicting the action mechanisms of toxic chemicals but also in evaluating their toxicity.

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