

Supplementary Information

Binding Kinetics Survey of the Drugged Kinome.

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EXPERIMENTAL PROCEDURES

Cell lines

HEK-293 cells (ATCC) or HeLa cells (ATCC) were cultured in DMEM (Gibco) + 10% FBS (GE Healthcare), with incubation in a humidified, 37°C/5% CO₂ incubator.

Plasmids

NanoLuc tagged kinase plasmids were from Promega. For general information see ¹³.

Proteins

Biotinylated kinases (Table S1) were purchased from Carna Biosciences. Except for His-tagged, biotinylated CDK2 and CDK9 which were synthesized in house according to previously described protocols ^{33,48}. In brief, human full-length CDK9 (N-terminal AviTagTM fusion protein) and cyclin T1 (N-terminal His₆ fusion protein) were co-expressed in Hi5 insect cells along with BirA for *in vivo* biotinylation according to the manufacturer (Avidity). Human full-length CDK2 (N-terminal His₆ and AviTagTM Fusion protein) was coexpressed in BL21(DE3) competent *E. coli* cells along with BirA. Both CDK9/cyclin T1 complexes and CDK2 were purified by immobilized metal ion affinity chromatography (HisTrap HP, GE-Healthcare).

Fluorescent probes and test compounds

Alexa Fluor 647-labeled Kinase Tracers 178, 199, 222, 236, 314 and 1710 (Table S2) were obtained from Thermo Fisher Scientific. Cell permeable Kinase Tracers K-4 and K-5 were from Promega. Kinase inhibitors (Table S4) were purchased from Selleck Chemicals.

Software and Databases

Analysis and visualization of kinetic screening data was performed using GraphPad Prism (V.6.00-7.00), Genedata Screener®, Biacore 4000 Evaluation Software and TIBCO Spotfire® (V.6.5.2.). Bioinformatics analyses were conducted with Clustal Omega (<https://www.ebi.ac.uk/Tools/msa/clustalo/>) and Python 3 (<http://www.python.org>). Chemoinformatics analyses were performed with TIBCO Spotfire (V.6.5.2.) and Pipeline Pilot (Version 16.5.0.143). QSP modeling and simulations were done with Copasi 4.1.9. and Phoenix version 6.4. Figures using a phylogenetic tree of the human kinome (reproduced courtesy of Cell Signaling Technology, Inc. (www.cellsignal.com)) were prepared using the software Kinome Render (<http://biophys.umontreal.ca/nrg/NRG/KinomeRender.html>) and KinMap (<http://kinhub.org/kinmap/>). Images of kinase-inhibitor cocrystal structures were drawn with the PyMOL Molecular Graphics System (Version 1.3). Ligand interaction diagrams were generated with Discovery Studio Visualizer (v17.1.0.16143).

Information about clinical phases of kinase inhibitors was retrieved from the 1st Oncology™ database. Kinase inhibitors plasma half-lives in healthy human adults were obtained from the IntegritySM database. Compounds' binding mode information (including DFG- and α C-helix conformations) and kinase-ligand interaction fingerprints were from KLIFS – a structural kinase-ligand interaction database⁴². Protein crystal structures complexed to inhibitors were downloaded from the Protein Data Bank (PDB).

Homogeneous time-resolved fluorescence resonance energy transfer (TR-FRET/HTRF) assays.

In general, TR-FRET (HTRF) assays were conducted as described in ³³. Experiments were performed at room temperature in black 384 Well Small VolumeTM HiBase Microplates (Greiner) using a buffer (50 mM Tris HCl, pH 7.5) containing 150 mM NaCl, 0.01% Tween20, 0.01% BSA and 2 mM DTT. Biotinylated kinases were incubated with 0.5 nM St-Tb (Streptavidin-Terbium (Cisbio Bioassay)) for labeling with the long-life fluorescent cryptate TR-FRET donor. Kinase Tracers 178, 199, 222, 236, 314 and 1710 served as fluorescent probes (TR-FRET acceptors). For all experiments, a nonspecific binding control (Ci) was performed in the presence of probe and 0.5 nM St-Tb, but in the absence of protein (= background signal). Fluorescence signals were detected in a PHERAstar (FS) plate reader (BMG LABTECH) by excitation of the donor with 5 laser flashes at 337 nm, followed by a time delay of 100 μ s and readout of acceptor and donor emission for 400 μ s at 665 and 620 nm respectively. HTRF signal (in the following often referred to as “probe binding signal”) was calculated from TR-FRET Ratio values as defined in the MARS data analysis software (BMG LABTECH):

$$\text{HTRF signal} = \text{TR-FRET Ratio} \times 10\,000 = [\text{acceptor counts} / \text{donor counts}] \times 10\,000$$

Development of equilibrium and kinetic probe competition assays (ePCA and kPCA). For assessment of the affinity of the fluorescent probe and for determination of the optimal amount of protein for the following TR-FRET experiments, an 8-point 2-fold serial dilution of the respective Kinase Tracer (volume: 2.5 μ L; final concentrations in assay: 3.125-400 nM) was incubated for at least 30 min with increasing concentrations of kinase (volume: 2.5 μ L; 7-point 2-fold serial dilution; final concentrations in assay: 0.125-8 nM). The minimal amount of protein for which the signal to background ratio is around 2.5-3.5 was selected for the following TR-FRET experiments, in order to obtain a sufficiently high signal at lowest

possible target protein concentration. Tracer-kinase pairs were selected following the recommendations by the supplier. In cases where no combinations were known for a particular kinase, the above experiments were performed using all Kinase Tracers.

The steady-state affinities of probe binding to the kinases were determined by removing the baseline (subtraction of nonspecific binding control signals) and fitting of the obtained probe-target binding signals to the “One site -- specific binding” model (GraphPad Prism):

$$Y = B_{\max} * X / (K_D + X)$$

Where X: concentration of probe (constrained as used in experiment); Y: specific binding signal; B_{\max} : maximum binding signal at very high probe concentrations; K_D : equilibrium dissociation constant in units of X

The kinetic rate constants of fluorescent probe binding to the individual kinases, as well as the optimal assay parameters with regard to probe concentration, kinetic time intervals and duration of observation were determined in association kinetics experiments. In a typical assay 5 μ L of target kinase (kinases concentrations as specified in Table S2) were added to an 8-point 2-fold serial dilution of probe (volume: 5 μ L; concentrations depended on determined affinity; two or more replicates) with the reagent injectors of the plate reader followed by immediate acquisition of TR-FRET signals. Prior to and after each usage, the injectors were washed with 1 M NaOH and ddH₂O. Specific signals from these association kinetics experiments were determined by removing the baseline (subtraction of nonspecific binding control signals) and globally fitted to the “Association kinetics -- two or more concentration of hot” model (GraphPad Prism):

$$K_D = k_{\text{off}} / k_{\text{on}}$$

$$k_{\text{obs}} = k_{\text{on}} * L + k_{\text{off}}$$

$$\text{Occupancy} = L / (L + K_D)$$

$$Y_{\text{max}} = \text{Occupancy} * B_{\text{max}}$$

$$Y = Y_{\text{max}} * (1 - \exp(-1 * k_{\text{obs}} * X))$$

Where K_D : calculated equilibrium dissociation constant [M], k_{obs} : observed rate of probe-target complex formation, k_{off} : dissociation rate constant of probe [s^{-1}], k_{on} : association rate constant of probe [$\text{M}^{-1}\text{s}^{-1}$], L : probe concentration [M]; Y is the specific binding signal; Y_{max} is the maximum observable signal at equilibrium; B_{max} : maximum binding signal at very high probe concentrations

Using Parameter constraints: L constrained to constant values as used in experiment; Shared parameters (between concentration traces) for global fit: k_{off} , k_{on} , B_{max}

Compound screening with ePCA. Analysis of steady-state affinities of the kinase inhibitors were performed in at least two independent experiments with two duplicates each. Compounds were diluted from 10 mM DMSO stocks in Storage 384 Microplates (REMP) using the PrecisionTM Microplate Pipetting System (BioTekTM). Serial dilutions (11-point 3.5-fold) were performed in 100% DMSO to yield 100-fold concentrated solutions that would result in final assay concentrations of 72.5 pM - 20 μM or 7.25 pM – 2 μM (depending on the inhibitor affinities) and 1% DMSO. To this end 50 nL of these stock solutions (or DMSO in control wells) were transferred from the stock plates into black 384 Well Small Volume HiBase Microplates assay plates (Greiner) using a Hummingbird Benchtop System (Digilab). The stock plates were sealed with adhesive aluminum foil by using the Quick Combi-Sealer (HJ-Bioanalytik) and stored at -20°C . The assay plates were either sealed with adhesive polypropylene foil and stored at -20°C or immediately used to run the ePCA: 3 μL of

fluorescent probe and 2 μ L of target kinase (see Table S2 for kinase specific assay parameters such as probe and target concentration) were added using a MultidropTM Combi Reagent Dispenser (Thermo Fisher Scientific) to start the competitive target kinase binding process of compound and probe. The assay was incubated for more than 1 h prior to acquisition of the TR-FRET signals. TR-FRET traces were normalized using high (100% probe binding) and low (0% probe binding = 100% inhibition) controls. These controls were calculated beforehand as mean of all signals in the respective control wells on a single microtiter plate (for 100% probe binding: wells with probe but without compound; 0% probe binding: wells with St-Tb and probe but without biotinylated kinase). Normalized data were fitted to a four-parameter sigmoidal dose-response curve using Genedata ScreenerTM with the proprietary automated data- and model driven outlier masking settings:

$$Y = \text{Bottom} + (\text{Top} - \text{Bottom}) / (1 + 10^{((\text{Log}(\text{IC}_{50}) - X) * \text{Hill slope}))}$$

Where Y: normalized probe binding signal (response); Bottom: bottom plateau (minimal response); Top: maximum response; X: logarithms of compound concentrations; IC_{50} : concentration of compound required to provoke a response half way between Bottom and Top (in units of compound concentration); Hill slope: slope factor describing steepness of dose-response curve

Parameter constrains: X constrained to constant values as used in experiment; Genedata Screener software was allowed to automatically choose if to fix or not to fix the Hill slope to 1, Bottom to 0 and Top to 100, but was not allowed to extrapolate IC_{50} s. Thereby fixing of the parameters was enforced if the data range is <30% or $r^2 < 0.1$; If not fixed, the parameters were constrained: Hill slope must be between 0.5 and 5; Bottom must be <30%; Top must be between 70 and 150%.

The IC₅₀ value was converted into the equilibrium dissociation constant (K_Deq) by using the Cheng-Prusoff equation ⁴⁹.

$$K_D = IC_{50} / (1 + L / K_{D_L})$$

Where K_D: equilibrium dissociation constant of compound; IC₅₀: concentration of compound required to provoke a response half way between minimum and maximum response; L: concentration of probe; K_{D_L}: equilibrium dissociation constant of probe

Compound screening with kPCA. Analysis of kinetic rate constants of inhibitor binding to kinases was performed in at least two independent experiments with two duplicates each. Compounds were diluted using the same instrumentation as for ePCA but with different volumes and dilution schemes. Briefly, 4-point 10-fold serial dilutions of 100x concentrated compounds were prepared in DMSO (final concentrations in assay: 2.5-2500 nM), and DMSO was applied into control wells (0% and 100% probe binding). 100 nL of these stock solutions were pre-dispensed to generate “assay-ready” plates, to which 5 µL of fluorescent probe were added prior to the experiment. To start the competitive binding assay, 5 µL of target kinase were added with the reagent injectors of the plate reader followed by immediate acquisition of TR-FRET signals (see Table S2 for kinase specific assay parameters such as probe and target concentration, measurement interval and observation time). For these experiments the script mode of the plate reader was used to dispense the target and measure plate octants as described in ³³. Prior to and after usage, the injectors were washed with 1 M NaOH and ddH₂O. To determine rate constants and affinity parameters, baseline signals (0% probe binding control, assumed to be linear with time) were subtracted from the kinetic traces

(including the zero compound concentration trace = 100% probe binding control) and globally fitted to the Motulsky-Mahan's⁵⁰ "Kinetics of competitive binding" equation:

$$Y(t) = \frac{B_{\max} \times K_1 \times L}{\text{Diff}} \left(\frac{K_4 \text{ Diff}}{K_F K_S} + \frac{K_4 - K_F}{K_F} e^{-K_F t} - \frac{K_4 - K_S}{K_S} e^{-K_S t} \right)$$

With

$$K_A = K_1 \times L + K_2$$

$$K_B = K_3 \times I + K_4$$

$$K_F = 0.5 \left(K_A + K_B + \sqrt{(K_A - K_B)^2 + 4 \times K_1 \times K_3 \times L \times I} \right)$$

$$K_S = 0.5 \left(K_A + K_B - \sqrt{(K_A - K_B)^2 + 4 \times K_1 \times K_3 \times L \times I} \right)$$

$$\text{Diff} = K_F - K_S$$

Where Y: specific probe binding signal; t: time [s]; K1: association rate of probe [$M^{-1}s^{-1}$]; K2: dissociation rate of probe [s^{-1}]; L: concentration of probe [M], K3: association rate of unlabeled compound [$M^{-1}s^{-1}$]; I: concentration of unlabeled compound [M]; K4: dissociation rate of unlabeled compound [s^{-1}]; B_{\max} : maximum binding signal at very high concentration of probe (and often exceeds what is measured in experiment)

Parameters constraints: K1, L, K2, I constrained to constant values as used in experiment and determined during assay development. Shared parameters (between concentration traces) for global fit: K3, K4, B_{\max} must be greater than 0

Affinities from kinetic analysis were calculated as the ratio of the on- and off-rate:

$$K_{D \text{ kin}} = k_{\text{off}} / k_{\text{on}}$$

Where $K_{D \text{ kin}}$: calculated equilibrium dissociation constant [M], k_{off} : dissociation rate constant [s^{-1}], k_{on} : association rate constant [$\text{M}^{-1}\text{s}^{-1}$]

Results from Monte Carlo analyses (Georgi et al. manuscript in preparation) were considered for evaluation of the data: If off-rates could not be determined accurately by nonlinear regression (if $\text{CV}(k_{\text{on}} \times K_{D \text{ eq}}) < 50\%$ and either case I) $\text{CV}(k_{\text{off}}) > 50\%$ or case II) k_{off} is not determinable by regression), off-rates were calculated using the equation as suggested in ³³:

$$k_{\text{off}} = K_{D \text{ eq}} \times k_{\text{on}}$$

Where $K_{D \text{ eq}}$: equilibrium dissociation constant [M] as determined in endpoint experiment ePCA, k_{off} : calculated dissociation rate constant in [s^{-1}], k_{on} : association rate constant [$\text{M}^{-1}\text{s}^{-1}$] as determined by using Motulsky-Mahan equation

Off-rates were set to > 0.35 (2 times faster than the fastest determined off-rate with a relative model error $< 40\%$), if $\text{CV}(K_{D \text{ kin}}) < 50\%$, the difference between $K_{D \text{ kin}}$ and $K_{D \text{ eq}}$ is less than five-fold and kinetic rate constants (k_{on} , k_{off}) are not determinable by regression (values far beyond assay quantitation limit and/or beyond diffusion limit: $k_{\text{on}} > 10^9 \text{ M}^{-1}\text{s}^{-1}$; $k_{\text{off}} > 10^4 \text{ s}^{-1}$ (case III); and/or $\text{CV} > 90\%$). Moreover, kinetic rate constants were considered as not evaluable, in all cases (which do not belong to case I-II but) with a more than five-fold difference between $K_{D \text{ kin}}$ and $K_{D \text{ eq}}$ (which could result from more complex interactions than simple 1:1 binding).

Residence times (RT) were calculated using the equation:

$$RT = 1/k_{\text{off}}$$

Where RT: residence time in units of time, k_{off} : dissociation rate constant in inverse time

Finally, the affinities determined with kPCA were corrected by considering the affinities determined with ePCA to derive a K_D (PCA): K_D (PCA) = $K_{D \text{ kin}}$, except for the following cases: (I) K_D (PCA) = $K_{D \text{ eq}}$, if $CV(K_{D \text{ eq}}) < 70\%$ and 1) $CV(K_{D \text{ kin}}) > 70\%$ or 2) $K_{D \text{ kin}}$ is n.e. or 3) reached assay quantitation limit or 4) there is at least a 5-fold discrepancy between $K_{D \text{ kin}}$ and $K_{D \text{ eq}}$; (II) else if $CV(K_{D \text{ kin}})$ and $CV(K_{D \text{ eq}}) > 70\%$, but there is a less than 5-fold discrepancy between $K_{D \text{ kin}}$ and $K_{D \text{ eq}}$, then K_D (PCA) = $\text{mean}(K_{D \text{ kin}}, K_{D \text{ eq}})$.

Cellular assay based on bioluminescence resonance energy transfer (BRET).

The cellular NanoBRET™ Target Engagement Assays (Promega) for ABL, BTK, FYN, KIT, LCK, Ret, SRC and Tie2 were conducted according to ¹³. Each kinase encoded a C-terminal fusion of NanoLuc luciferase with the exception of ABL, for which NanoLuc was at the N-terminus. In brief, cells were transiently transfected with NanoLuc / kinase fusion plasmids using FuGENE HD according to the manufacturer's protocol (Promega) and seeded at a density of 2×10^5 cells per mL into white, sterile 96-well plates (Corning). 24 hours post-transfection, growth medium (90% DMEM + 10% fetal bovine serum) was replaced with Assay Medium (Opti-MEM® I Reduced Serum Medium, no phenol red). Then, 1 $\mu\text{L/mL}$ of

1000x concentrated test compound was added (final concentrations: [Staurosporine]: 0.5 μ M; [Imatinib]: 2 μ M; [Ibrutinib], [Dasatinib], [Ponatinib]: 0.2 μ M) and incubated at 37°C for 3 h. After removing unbound test compound via washout with prewarmed Assay Medium and a wash step, 100 μ L Assay Medium and 1 μ M (final concentration) cell-permeable fluorescent Kinase Tracer K-4 or KinaseTracer K-5 (Promega) were added, followed by immediate addition of the Complete NanoBRET™ NanoGlo® Substrate / Extracellular NanoLuc® Inhibitor Solution according to the manufacturer's protocol (Promega). Subsequently, bioluminescence resonance energy transfer was measured over 5 h in 3-minutes intervals with a GloMax® Discover Instrument equipped with NanoBRET filters. A positive control with saturating dose of kinase inhibitor (0% probe binding) served for background correction, and a probe signal control (100% probe binding) was conducted by adding only DMSO instead of compound solubilized in DMSO. For better comparability with simulated target occupancies, NanoBRET™ traces were normalized to obtain binding signals as percentage of control, according to the equation:

$$\text{Signal}_{\text{normalized}}(t) = (1 - (\text{Signal}(t) - \text{Signal}_{0\%}(t)) / (\text{Signal}_{100\%}(t) - \text{Signal}_{0\%}(t))) \times 100\%$$

t: time; $\text{Signal}_{0\%}(t)$: 0% probe binding signal; $\text{Signal}_{100\%}(t)$: 100% probe binding signal

Changes in cellular target occupancy (Δ TO) were calculated as follows:

$$\Delta\text{TO}_{0-t \text{ min}} [\%] = 100 - (\text{Signal}_{\text{normalized}}(t) / \text{Signal}_{\text{normalized}}(t=0)) * 100$$

Where t: time; $\text{Signal}_{\text{normalized}}(t)$: normalized signal at time = t; $\text{Signal}_{\text{normalized}}(t=0)$: normalized signal at time = 0

Constrains for experimental traces: $\text{Signal}_{\text{normalized}} (t=0)$ was estimated based on fitting of the nomalized trace to “One phase decay” curve using GraphPad Prism. $\Delta\text{TO}_{0-t \text{ min}}$ were set to 100% for all values >100% and to 0% for all values <0%.

Changes in simulated cellular target occupancy (see below) were calculated in the same way, but without consideration of the above mentioned constrains.

QUANTIFICATION AND STATISTICAL METHODS

General statistics.

For assessment of the direction and strength of relationship between k_{on} , k_{off} , K_D values determined in different assays or replicates and their relationship with changes in cellular target occupancy (ΔTO) (using GraphPad Prism), Spearman correlation calculations were utilized to compute correlation coefficients and two-tailed P values (binding parameters are approximately log normally distributed, but skewed due to assay quantitation limits). Agreement between methods and replicates was evaluated by Bland-Altman analysis using mean log differences. Statistical analyses were performed with log values of binding parameters, including generation of histograms (bin: 0.5) and distance matrices calculations, Gini score calculation and multivariable regression analyses (for more details see below). Fitting of log values of binding parameters to a “Gaussian” curve or “Cumulative Gaussian -- Percents” curve, respectively, as well as descriptive statistics was performed using GraphPad Prism. Percent differences were calculated as defined here:

$$\text{Percent difference} = (\text{value1} - \text{value2}) / (\text{value1} + \text{value2}) * 100$$

Gini scores as metric for selectivity.

Gini scores were determined based on normalized log values of binding parameters as described by Graczyk et al.³⁶. Since Gini score calculation requires values normalized between 0 and 100, log values were normalized based on minimum and maximum values.

Plasma and dissociation half-lives.

Plasma half-lives of the kinase inhibitors (applied for Fig. 3a) were calculated as median (with 25% and 75% quantil) of all entries (for our tested KI) in IntegritySM database for healthy human adults. The median of these plasma half-lives was used to calculate an average elimination rate k_{el} (applied for Fig. 3b): $k_{el} = \ln(2)/\text{plasma-half-life}$

Quantitative systems pharmacology and pharmacokinetics simulations.

Model equations for simulations of target occupancy *in vivo*

All simulations were performed until *in vivo* target occupancy traces reached equilibrium (same profile will be reached in next dosing interval).

Pharmacokinetics-binding kinetics (PK-BK) model

To simulate the compound-concentration-time profile after oral administration of a kinase inhibitor, a two-compartment model was used in Copasi 4.1.9. A simple 1:1 drug-target binding model, assuming the free drug hypotheses and constant total protein target concentrations, was added to describe compound binding to protein targets. This PK-BK model for compound binding to n proteins is given by the following equations:

PK-BK model

$$\frac{d([C_1](t))}{dt} = \frac{k_a \times A_a(t)}{V_1} - CL \times [C_1](t) - CLD \times ([C_1](t) - [C_2](t)) + \sum_{i=1}^n (k_{off_i} \times [TC]_i(t) - k_{on_i} \times (fu \times [C_1](t)) \times [T]_i(t))$$

$$\frac{d([TC]_i(t))}{dt} = k_{on_i} \times (fu \times [C_1](t)) \times [T]_i(t) - k_{off_i} \times [TC]_i(t) \quad \text{for } i = 1, 2, \dots, n$$

$$\frac{d([C_2](t))}{dt} = CLD \times ([C_1](t) - [C_2](t))$$

$$\frac{d(A_a(t))}{dt} = -k_a \times A_a(t)$$

Target occupancies can be calculated using $TO_i(t) = [TC]_i(t) / [T_{total}]_i \times 100\%$ for $i = 1, 2, \dots, n$

Where n is the number of target kinases. The total target concentrations $[T_{total}]$ are equal to the sum of the respective free protein $[T]$ and inhibitor-complexed protein $[TC]$ concentrations. k_{on} and k_{off} represent the rate constants for association and dissociation of the compound-target complex. $[C_1]$ ($= A_1/V_1$) and $[C_2]$ ($= A_2/V_2$) are the total compound concentrations ($=$ amount/distribution volume) in the central and peripheral compartment, respectively. A_a is the amount of drug in the gastrointestinal tract (effective dose $=$ dose \times bioavailability F). fu is the unbound fraction of the compound in plasma. k_a , CL , CLD represent the rate constants for absorption, elimination and distribution.

The following assumptions and initial values were used for the simulations:

The total target concentrations $[T_{total}]$ were fixed to 10 nM (within the lower range of a typical concentration of signaling protein according to Milo et al. BioNumbers database⁵¹ and as reported in literature and assumed in previous analyses²³); $[TC](t=0) = 0$; $A_1(t=0) = A_2(t=0) = 0$; The model parameters k_a , V_1 , V_2 , CL and CLD were estimated on the

basis of the concentration-time profiles (using Phoenix version 6.4, Pharsight), F was fixed to 1. Administered dose, dosing interval, fu and concentration-time profiles are given in Clinical Pharmacology and Biopharmaceutics Review documents for ibrutinib (NDA: 205552) and dasatinib (NDA: 21-986; 22-072).

Pharmacokinetics and the concentration-affinity effect – PK-K_D model

Steady state affinity driven compound-concentration-time profiles were simulated with the two-compartment model as described above, but without consideration of the drug-target binding model terms and the target occupancy traces were calculated using:

$$TO(t) = \frac{f_u \times [C_1](t)}{K_D + f_u \times [C_1](t)} \times 100\%.$$

Model equations for simulations of target occupancy in cells

Binding Kinetics - Permeation model

To simulate cellular target occupancy profiles in Copasi 4.1.9, a Binding Kinetics (BK) - Permeation model was developed. A simple 1:1 drug-target binding model with competition between compound and ATP was used to describe compound binding to protein targets. The total protein target concentration was assumed to be constant. Permeation of the compound through the cellular membrane was approximated using Fick's First Law. The BK-Permeation model is given by the following equations:

BK-Permeation model

$$\frac{d([C_1](t))}{dt} = k_{\text{off}} \times [TC](t) - k_{\text{on}} \times [C_1](t) \times [T](t) - P \times A_{\text{O,cell}} \times ([C_1](t) - [C_2](t)) / V_1$$

$$\frac{d([TC](t))}{dt} = k_{\text{on}} \times [C_1](t) \times [T](t) - k_{\text{off}} \times [TC](t)$$

$$\frac{d([T](t))}{dt} = k_{\text{off}} \times [TC](t) - k_{\text{on}} \times [C_1](t) \times [T](t) + k_{\text{off,ATP}} \times [T.ATP](t) - k_{\text{on,ATP}} \times [ATP](t) \times [T](t)$$

$$\frac{d([ATP](t))}{dt} = k_{\text{off,ATP}} \times [T.ATP](t) - k_{\text{on,ATP}} \times [ATP](t) \times [T](t)$$

$$\frac{d([T.ATP](t))}{dt} = k_{\text{on,ATP}} \times [ATP](t) \times [T](t) - k_{\text{off,ATP}} \times [T.ATP](t)$$

$$\frac{d([A_{2l}](t))}{dt} = P \times A_{\text{O,cell}} \times ([C_1](t) - [C_2](t))$$

Target occupancy can be calculated using $TO(t) = [TC](t) / [T_{\text{total}}] \times 100\%$

The total concentration of target protein $[T_{\text{total}}]$ is equal to the sum of free protein $[T]$ and inhibitor-complexed protein $[TC]$ and ATP-complexed protein $[T.ATP]$ concentrations. k_{on} and k_{off} represent the rate constants for association and dissociation of the compound-target complex. $[ATP]$ and $[T.ATP]$ are the concentrations of ATP and ATP-complexed protein. $k_{\text{on,ATP}}$ and $k_{\text{off,ATP}}$ represent the rate constants for association and dissociation of the ATP-target complex. $[C_1](=A_1/V_1)$ and $[C_2](=A_2/V_2)$ are the compound concentrations (= amount/volume) within cells and outside cells, respectively. P is the permeability coefficient of compound through membrane. $A_{\text{O,cell}}$ is the total surface area of the cells.

The following assumptions and initial values were used for the simulations:

For simulation of the pre-incubation step (TO onset) (3h): $A_{O,cell} = 0.0154 \text{ dm}^2$ and $V_1 = 4 \times 10^{-7} \text{ L}$ (cumulated sphere surface area and volume of 2×10^5 cells); $V_2 = 0.001 \text{ L}$ (2×10^5 cells / mL were applied in the nanoBRET preincubation step); $[TC](t=0) = 0$; $[T.ATP](t=0) = [ATP]/([ATP] + K_{D,ATP}) \times [T_{total}]$ (equilibrated ATP-target complex); $[T](t=0) = [T_{total}] - [T.ATP]$; $A_1(t=0) = 0$; $A_2(t=0)$ is the compound concentration applied in the experimental pre-incubation step.

For simulation of target occupancy decline after washout: $[T]$, $[TC]$, $[T.ATP]$, $[C_1]$ and $[ATP]$ at $t=0$ were set to the output parameters of preincubation simulations; $A_2(t=0) = 0$ (washout); $A_{O,cell} = 0.0014 \text{ dm}^2$ and $V_1 = 3.6 \times 10^{-8} \text{ L}$ (cumulated sphere surface area and volume of 18000 cells (90 μL of 2×10^5 cells / mL were applied in the washout experiment)); $V_2 = 0.0002 \text{ L}$ (applied assay volume).

In both parts of the simulation, the total concentration of the overexpressed target protein $[T_{total}]$ was fixed to 100 nM. ATP concentration was fixed to 0.0001 M (lower than typical cellular ATP concentrations as dead cells lose ATP⁵²) or 0.001 M, respectively. ATP k_{on} and k_{off} were assumed to be similar for the studied kinases and were fixed to $9.2 \times 10^5 \text{ M}^{-1}\text{s}^{-1}$ and 1.9 s^{-1} ⁵³. Permeability coefficients (P) were set to 10^{-6} cm/s ⁵⁴ initially, but were reduced for Imatinib (10^{-7}) and Ponatinib (10^{-8}) to simulate additional effects reducing compound intake and release. A cell size of $2000 \mu\text{m}^3$ in volume (typical size according to Milo et al. BioNumbers database⁵¹) was assumed for calculation of the cumulated surface area and volume of the cells. For simulation of covalent binding of ibrutinib, the off-rate was set to 10^{-7} s^{-1} (RT of 116 days).

Steady state affinity driven target occupancy profiles were simulated using a K_D -Permeation model.

K_D- Permeation model

The compound-concentration-time profile was simulated by:

$$\frac{d([C_1](t))}{dt} = -P \times A_{O,cell} \times ([C_1](t) - [C_2](t))/V_1$$

$$\text{and } \frac{d([C_2](t))}{dt} = P \times A_{O,cell} \times ([C_1](t) - [C_2](t))/V_2$$

and target occupancy traces were calculated using

$$TO(t) = \frac{[C_1](t)}{K_D \left(1 + \frac{[ATP](t)}{K_{D,ATP}}\right) + [C_1](t)} \times 100\%$$

Where K_D represents the equilibrium dissociation constant of the compound-target complex. [ATP] and is the concentration of ATP. K_{D,ATP} represents the equilibrium dissociation constant of the ATP-target complex. [C₁] (= A₁/V₁) and [C₂] (= A₂/V₂) are the compound concentrations (= amount/volume) within cells and outside cells, respectively. P is the permeability coefficient of compound through membrane. A_{O,cell} is the total surface area of the cells.

Otherwise, the simulations (of pre-incubation step and target occupancy decline after washout) were conducted by using the same assumptions and initial values as described above (for the BK-Permeation model).

Distance matrices and hierarchical clustering.

Distance matrices were calculated as ‘100 minus percent similarity’.

Compound structure similarity matrix was generated with Pipeline Pilot (Version 16.5.0.143) using the component ‘Fingerprint Similarity NxN’ (ECFP4 fingerprint; distance measure: Tanimoto). Multiple sequence alignment of kinase constructs was performed using Clustal Omega ⁵⁵.

Compound and kinase similarities were calculated with Python 3 (available at <http://www.python.org>) and the module SciPY (scipy.spatial.distance using the distance measure ‘Euclidean’) based on vectors of log values of binding parameters (pK_D, -pk_{on}, pk_{off}). No scaling or normalization was applied, on-rates and off-rates for kinase-inhibitor pairs with affinities below assay quantitation limit were set to 350 M⁻¹s⁻¹ and 3.5 s⁻¹, ‘not evaluable’ values were masked. In detail, Euclidean distances were calculated as follows:

$$d(f_j, f_k) = \sqrt{\sum_{i=1}^n (f_{j_i} - f_{k_i})^2}$$

Where for the compound-centric distance matrices: f is a feature for kinase-compound binding (pK_D, -pk_{on} or pk_{off}, respectively); j is the jth compound; k is the kth compound; i is the ith kinase and n is the total number of kinases;

and where for the kinase-centric distance matrices: f is a feature for kinase-compound binding (pK_D, -pk_{on} or pk_{off}, respectively); j is the jth kinase; k is the kth kinase; i is the ith compound and n is the total number of compounds.

Clustering/Visualization of distance matrices was performed with Python 3 using the modules Pandas, Seaborn and SciPY (scipy.spatial, scipy.cluster.hierarchy (routine linkage for

agglomerative clustering based on clustering method 'average')). Pearson correlation coefficients were used as measure for correlation between distance matrices.

Definition and calculation of physicochemical properties of compounds.

The molecular properties and chemical functionalities of the compounds were used as defined here: The molecular weight (MolWt) of the compounds was calculated based on the molecular formula and assuming standard isotope distribution. Flexibility is the number of rotatable bonds divided by number of bonds (between heavy atoms). Rotatable bonds are the single bonds between heavy atoms that are both not in a ring and not terminal (connected to a heavy atom which is attached to only hydrogens) as well as no amide C-N bonds. F, Cl, Br, and I atoms in a compound account for the number of halogens. The sp^3 C / C ratio is the number of sp^3 carbon atoms divided by the total number of carbon atoms. Hydrogen bond donors (HBD) are the heteroatoms (O,N,P,S) with one or more attached hydrogen atoms, while hydrogen bond acceptors (HBA) are heteroatoms (O,N,P,S) with one or more lone pairs, excluding atoms with positive formal charges, amide and pyrrole-type nitrogens, and aromatic oxygen and sulfur atoms in heterocyclic rings. The polar surface area was calculated from the molecular topology, considering all N and O atoms including attached H atoms according to Ertl⁵⁶, but excluding sulfur and phosphorus polar fragments. The prediction of predominant charge state (predominantly positive, negative or neutral molecule, based on pKa prediction) relied on Simulations Plus' ADMET predictor (v.1, 2013.0620) software. The property was applied as a qualitative variable in the analyses below and set to 1 (positive), 0 (neutral) or -1 (negative). Likewise: The compound's binding mode information (involved subpockets and conformations, including DFG- and α C-helix conformations) and kinase-ligand interaction fingerprints (obtained from KLIFS database) are qualitative descriptors: 1

or 0 representing whether or not the kinase ligand interaction shows the respective binding feature.

Multivariable linear regression based on a genetic function approximation algorithm.

Multivariable linear regression models for log values of binding parameters (as dependent variable) were built with a genetic function approximation algorithm using molecular descriptors (e.g. number of HBD and HBA, $\text{sp}^3\text{C} / \text{C}$) and/or structural features (kinase-compound interactions or binding mode information) as predictors. This was computed with Pipeline Pilot (Version 16.5.0.143) using the component ‘Learn Molecular GFA Model’ (for details of GFA algorithm see reference ³⁹) with the following parameters: population size = 100, maximum generations: 1000; initial equation length: 4, maximum equation length = 15, Scoring method: Pareto (non-dominated sorting genetic algorithm II) with Score Function: Adjusted R-squared (estimates most appropriate number of variables and penalizes overfitting).

Additionally, correlation matrices were generated with Pipeline Pilot (Version 16.5.0.143) using the component ‘Correlation Matrix’ calculating Pearson correlation coefficients.

SUPPLEMENTARY FIGURES

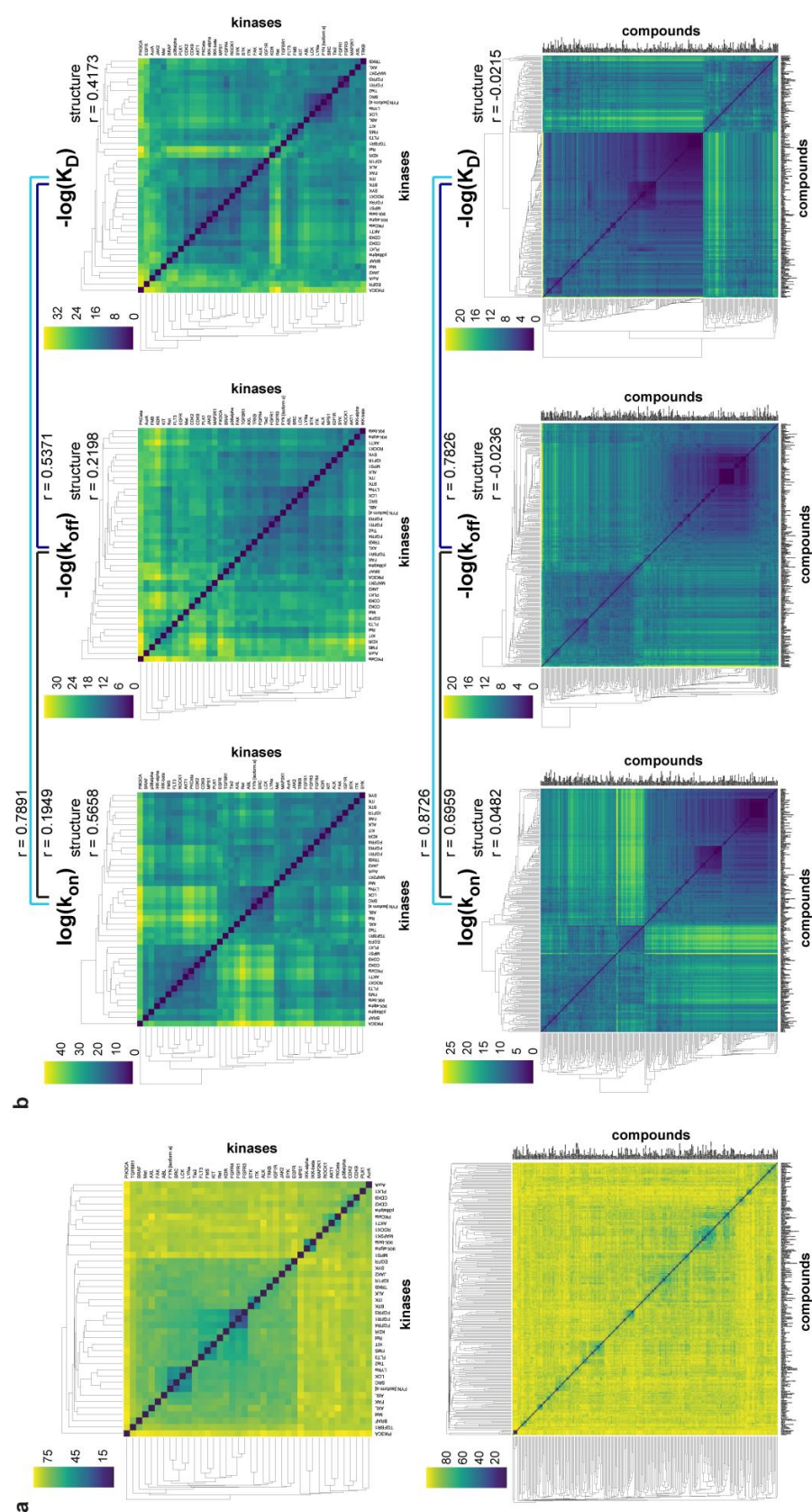


Figure S1. Related to Figures 1 and 5. Distance matrices based on structural similarities, affinity and BK parameters per kinase (kinase-centric) and kinase inhibitor KI (compound-centric). (a) Structural similarities of kinases and inhibitors investigated: Clustered Euclidean distance matrix for kinases based on sequence similarities were calculated with Clustal Omega. Clustered distance matrix for compounds are based on Tanimoto similarities of structure ECFP4 fingerprints. (b) Clustered distance matrices for affinity and BK parameters per kinase and KI: The colors indicate the Euclidian distances and the r values the correlation between these matrices and their correlation with the kinase sequence homology- and compound structure fingerprint-based distance matrices shown in panel a.

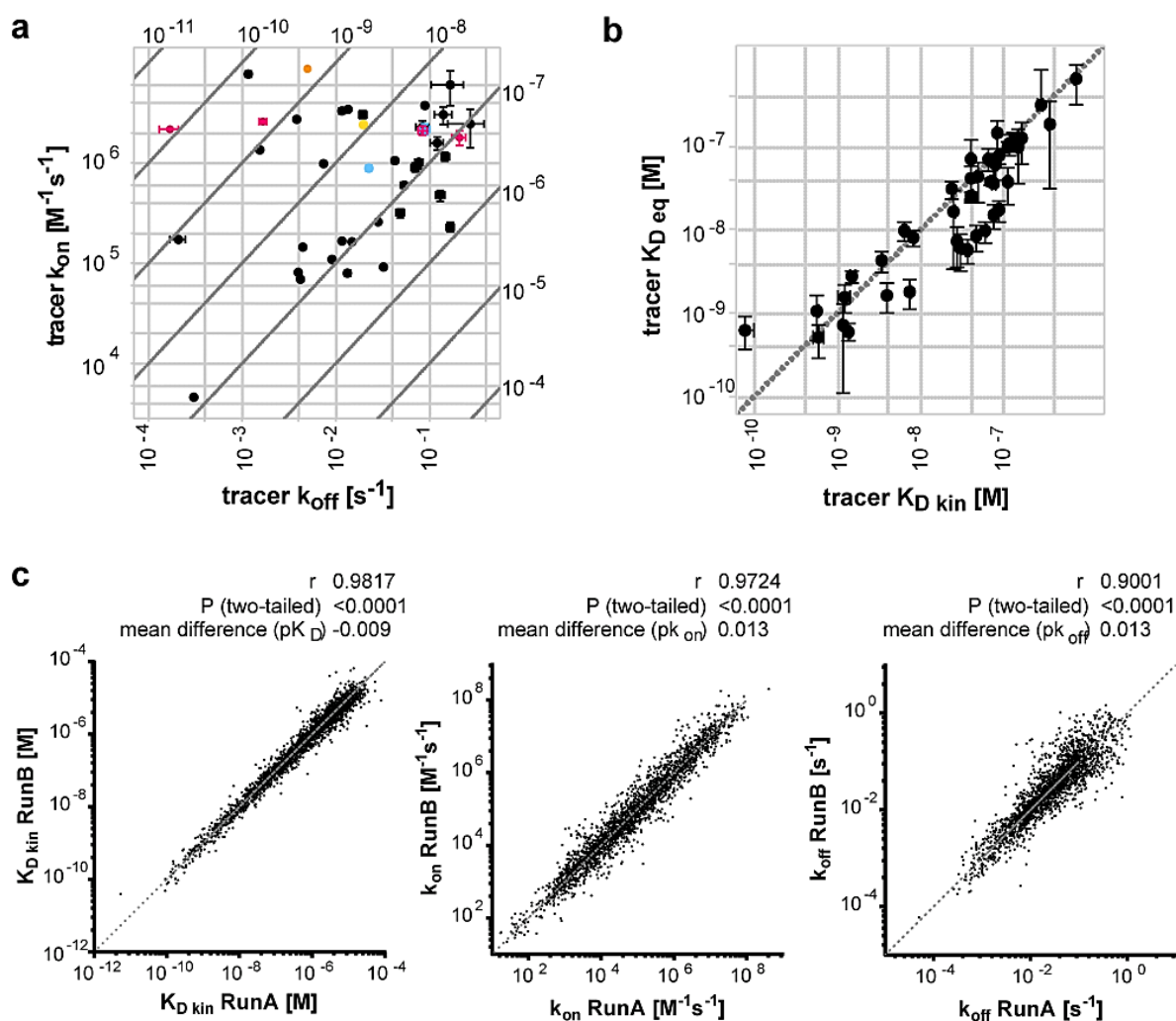


Figure S2. Related to Figure 1. Validation of assay panel for BK profiling of KIs. (a) Binding properties of Kinase Tracers validated in this study: RaPID plot showing BK and affinity parameters extracted from kinetic association experiments (Table S3) of the Kinase Tracers binding to 40 kinases (color code: black - Kinase Tracer 236; pink - Kinase Tracer 178; blue - Kinase Tracer 199; yellow - Kinase Tracer 222; orange - Kinase Tracer 314). (b) Comparison of affinities obtained from kinetic association experiments and steady state titrations of the probes. (c) Reproducibility of KI BK measurements performed in this study: Comparison of affinities (left panel), on-rates (middle) and off-rates (right panel) obtained in two independent runs of kPCA experiments (Run A and Run B). Statistical parameters from Spearman and Bland-Altman analyses are indicated in the graphs.

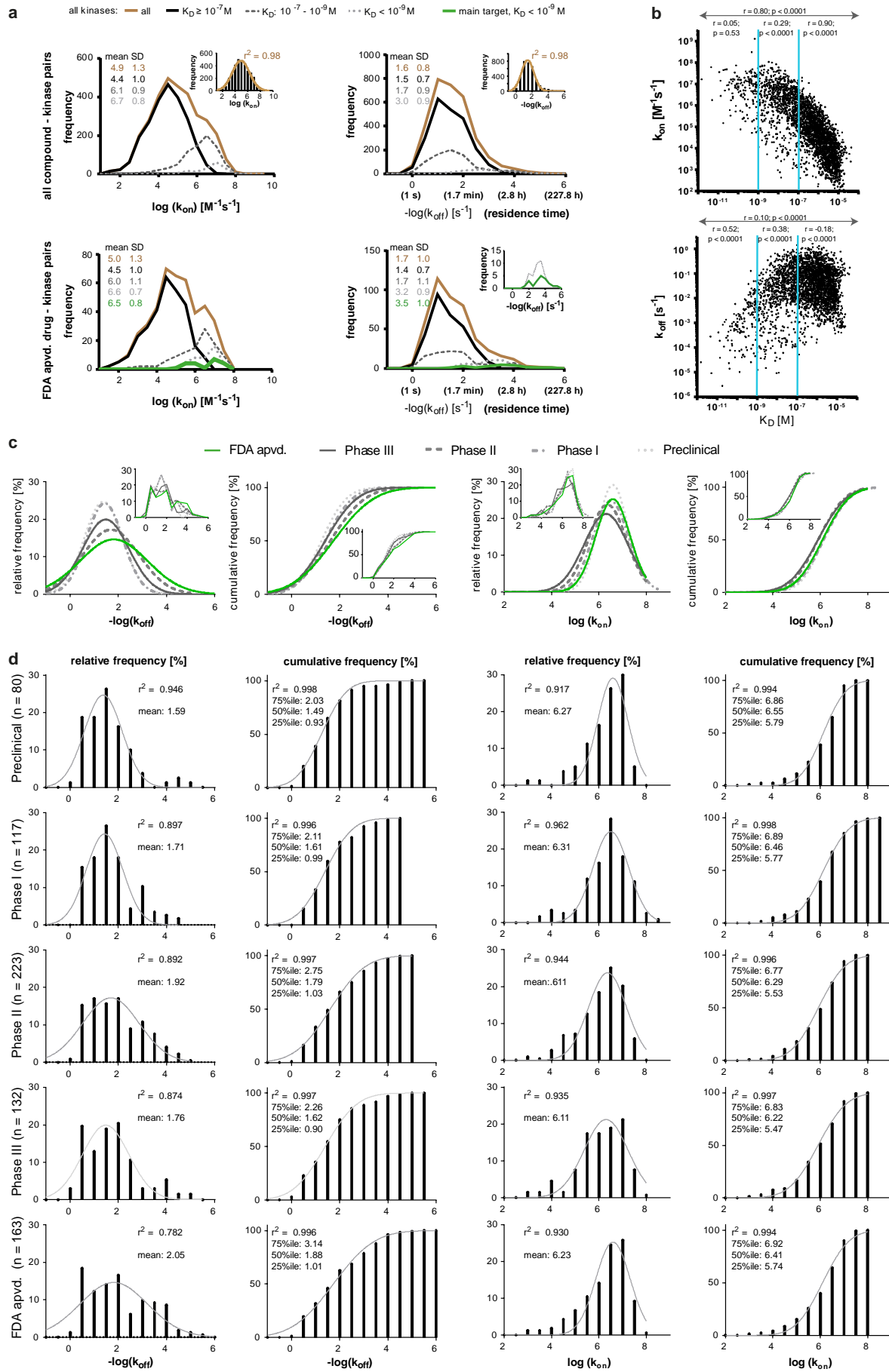


Figure S3. Related to Figure 2. Distribution of BK parameters quantified in this study (3230) and its dependence on the clinical phase of the KI. (a) Histograms (Frequency = No. of values) showing on- and off-rates distributions (left and right panels respectively) depending on the affinities of all kinase-inhibitor pairs (top) and those approved by the FDA (bottom). To ease visualization, bins are omitted and histograms are displayed as curves connecting frequency values of adjacent bins. The insets in the top panels represent histograms for all kinase-inhibitor pairs and the corresponding fits to the Gaussian distribution (as solid lines). (b) Comparison of all on- and off-rates (top and bottom panels respectively) to the compound affinities measured. Only parameters within assay detection limits are shown. Spearman correlation statistics for the complete dataset and three affinity groups (indicated by cyan lines) are indicated on top of the graphs. (c-d) Relative and cumulative frequency distributions of rate constants among KIs' from different phases in the pharmaceutical development pipeline: Distribution histograms corresponding to off-rates (first two panels from left) and on-rates (third and fourth panels from left) for all interactions with $K_D < 10^{-7}$ M grouped by inhibitors' clinical development status. Relative frequency distributions are shown in the first and third panels and cumulative frequency distributions in the second and fourth panels. Panel (c) shows all fits to the Gaussian distribution. The corresponding raw data histograms are shown as insets (To ease visualization, bins are omitted and histograms are displayed as curves connecting frequency values of adjacent bins). (d) Histograms (bars) and fits to the Gaussian distribution (solid lines) for the compounds whose BK could be quantified in this study. Descriptive statistics values are indicated in the graphs.

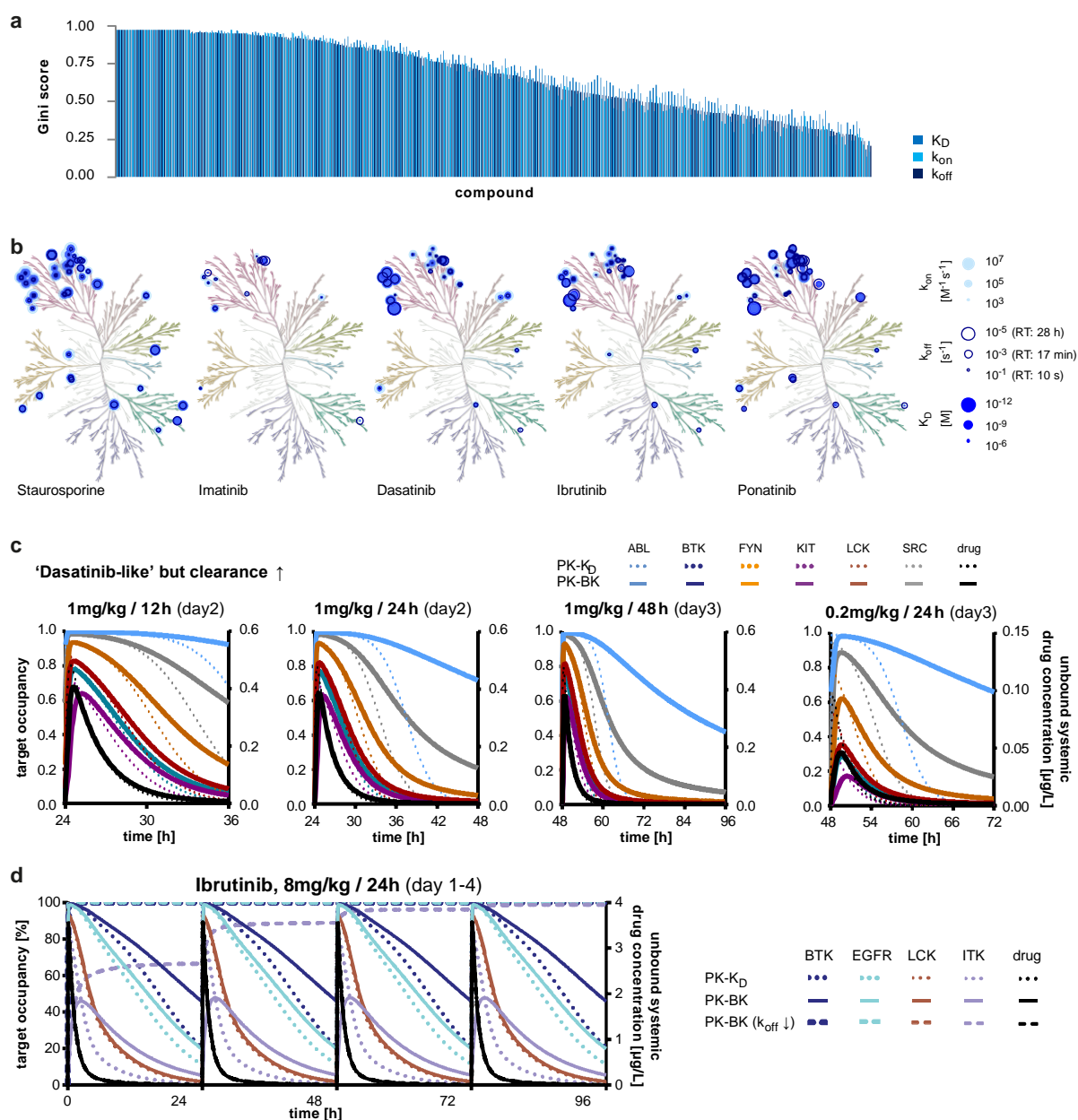


Figure S4. Related to Figure 3. Assessment of KI kinetic selectivity with classic vs. new approaches. (a) Overall selectivity of KIs evaluated in this study (with methods designed to analyze steady state binding data): Gini selectivity scores of all KIs based on their affinities, on-rates or off-rates. (b) Phylogenetic trees representing the human kinome and the binding of five KIs. The symbol colors represent the individual binding parameters measured and their size their magnitude. (c-d) QSP simulations of dynamic target occupancy by KI *in vivo*: (c) Simulations of the time course of *in vivo* TO by a hypothetical 'Dasatinib-like' compound with a faster clearance (4-fold intercompartmental clearance; 10-fold plasma clearance) but otherwise identical BK and PK parameters as the KI Dasatinib, from the binding kinetics (PK-BK) and affinity (PK- K_D) perspectives. Longer dosing intervals lead to higher discrepancies between the models (indicating hysteresis), but also to lower target occupancies at the end of the dosing interval in the PK-BK model. A lower dose (right panel) leads to lower maximum target occupancies, especially in the PK-BK model. (Day: day where target occupancy traces reached equilibrium). (d) Ibrutinib's main target BTK and three off-targets (EGFR, LCK and ITK) were followed using three models: one in which TO is driven by steady state affinity

(PK- K_D), a second one assuming BK for a 1:1 reversible interaction (PK-BK), and a third one considering covalent inhibition of BTK, EGFR and ITK (PK-BK ($k_{off}\downarrow$)). Current clinical dosage and administration schedules were used.

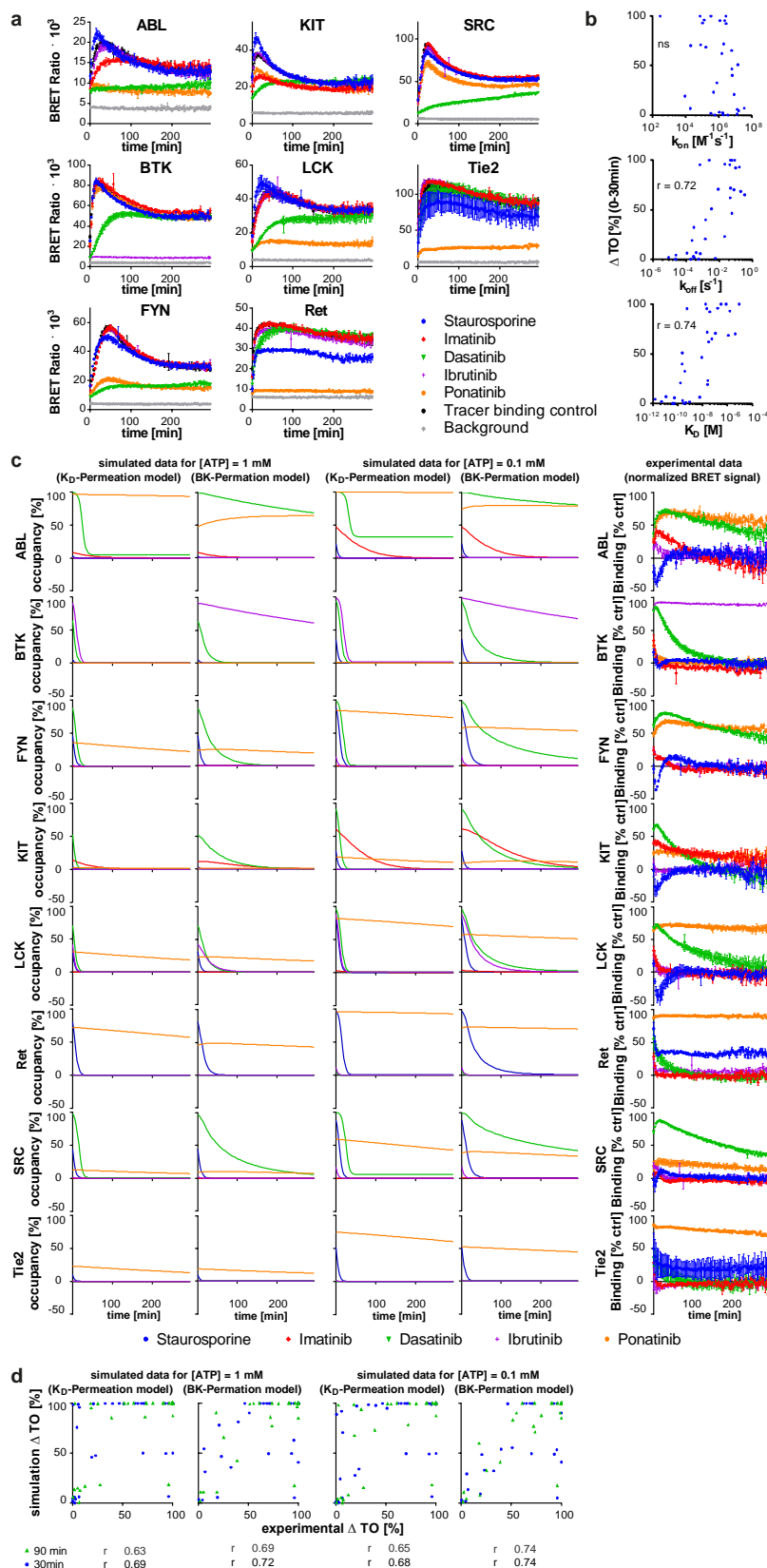


Figure S5. Related to Figure 4. Translation of biochemical BK parameters to dynamic TO in cells (full dataset).

(a) NanoBRET™ assays for dynamic target engagement by KIs in intact cells: Time courses of NanoBRET™ probe binding to eight RTKs in intact cells after treatment with six KIs or DMSO, followed by drug washout.

(b) Comparison of the on-rates, off-rates or affinities for kinase-inhibitor pairs with percent changes in TO (Δ TO) during the first 30 minutes after washout (as measure for intracellular target engagement).

Correlation coefficients are indicated in the graphs (n.s. if analysis not significant). (c) Simulated time course of TO decline upon compound washout using equilibrium affinity (K_D)-based (column 1, 3) or BK-based (column 2, 4) permeation models and either assuming 1 mM ATP (column 1-2) or 0.1 mM ATP (column 3-4), compared to normalized experimental data from NanoBRET™ washout experiments (right column) shown in panel a. (d) Comparison between simulated and experimental traces by using percent changes in TO (Δ TO) during the first 30 and 90 minutes after washout.

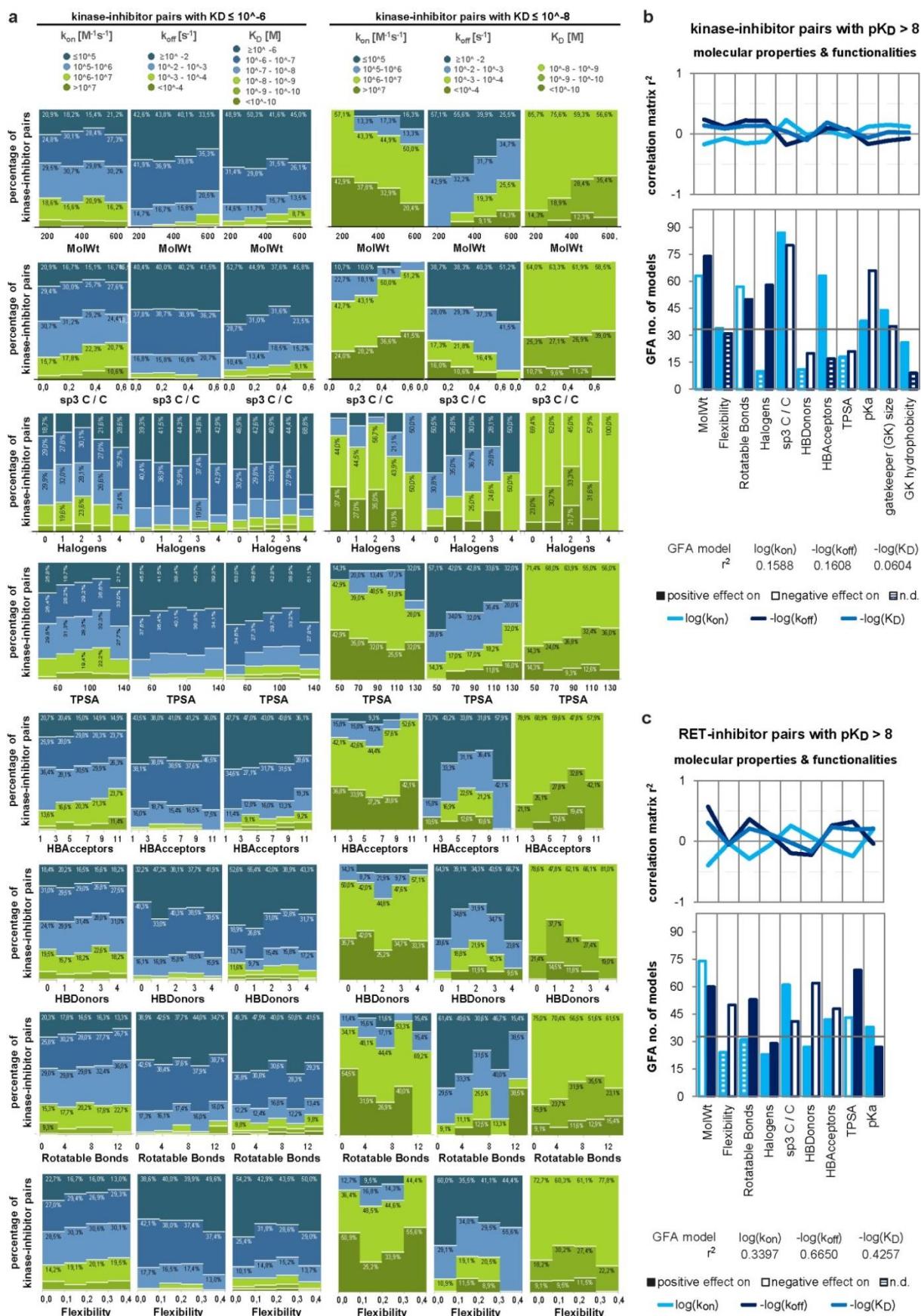


Figure S6. Related to Figure 5. Compound properties influencing affinity and BK parameters. (a) The bar charts represent relative (percentage) on-, off-rates and affinity distributions for KI pairs with $K_D \leq 10^{-6}$ M (left panel) or $K_D \leq 10^{-8}$ M (right panel). Each row show a different compound molecular property (b) Relationships between BK parameters and molecular properties (for kinase-inhibitor pairs with $K_D \leq 10^{-8}$ M) found with genetic function approximation (GFA). Line plots and bar charts represent exemplary results of correlation analyses (r^2) and multivariable linear regression models built (number of variable usage in GFA models) respectively. (c) Same analysis was conducted for the RTK Ret only. The higher r^2 values support the potential utility of this type of analysis in SKR studies focusing on a single target.

Figure S7. Related to Figure 6. Effect of KI pose and structural features of kinase-inhibitor complexes on affinity and BK parameters. (a) The bar charts represent relative (percentage) on-, off-rates and affinity distributions for kinase-inhibitor pairs with $K_D \leq 10^{-6}$ M or $K_D \leq 10^{-8}$ M. Bottom and top panels show different KI pose (i.e. kinase conformation as specified in KLIFS database for kinase-inhibitor pairs with PDB entry). (b) 85 binding site residues specified in the KLIFS database were investigated using correlation analyses and multivariable linear regression models built by genetic function approximation (considering only kinase-inhibitor pairs with PDB and KLIFS entry). The goal was to explore the relationships between specific interactions of the compounds and their binding properties (k_{on} , k_{off} , K_D). Correlation matrix r^2 values are shown as line plot (left), number of variable usage in models built by GFA are represented as bar chart (middle), and mean GFA coefficients for individual binding parameters are displayed as heatmap (right). The graphs show all interactions which are different for at least two kinase-inhibitor pairs.

SUPPLEMENTARY TABLES

Table S1. Related to Figures 1 and S1: Kinase constructs used in this study

Kinase	Gene ID	UniProt	Amino acids	Main target of FDA-approved drug screened in this study after ¹
ABL	25	P00519	Full-length 2-1130	Bosutinib, Dasatinib, Imatinib, Nilotinib, Ponatinib
AKT1	207	P31749	catalytic domain 104-480	
ALK	238	Q9UM73	cytoplasmic domain 1058-1620	CH5424802 (Alectinib), Crizotinib
AurA	6790	O14965	Full-length 1-403	
AXL	558	P30530	cytoplasmic domain 464-885	
BRAF	673	P15056	catalytic domain 433-726	Dabrafenib, Vemurafenib
BTK*	695	Q06187	Full-length 2-659	Ibrutinib
CDK2	1017	P24941	Full-length 1-298	
CDK9	1025	P50750	Full-length	
CHK1	1111	O14757	Full-length 1-476	
EGFR	1956	P00533	cytoplasmic domain 669-1210	Afatinib, Vargatef (Nintedanib), Desmethyl Erlotinib, Gefitinib, Lapatinib, Vandetanib
FAK	5747	Q05397	Truncated 376-1052	
FGFR1*	2260	P11362	cytoplasmic domain 398-822	
FGFR3*	2261	P22607	cytoplasmic domain 436-806	
FGFR4	2264	P22455	cytoplasmic domain 460-802	
FLT3	2322	P36888	cytoplasmic domain 564-993	
FMS	1436	P07333	cytoplasmic domain 538-972	
FYN [isoform a]	2534	P06241	Full-length 1-537	
IGF1R*	3480	P08069	cytoplasmic domain 959-1367	
IKK-alpha	1147	O15111	Full-length 1-745	
IKK-beta	3551	O14920	Truncated 1-662	
ITK*	3702	Q08881	Full-length 2-620	
JAK2	3717	O60674	catalytic domain 826-1132	Ruxolitinib, Tofacitinib
KDR	3791	P35968	cytoplasmic domain 790-1356	Apatinib, Axitinib, Vargatef (Nintedanib), Lenvatinib, Pazopanib, Sunitinib, Vandetanib, Cabozantinib
KIT*	3815	P10721	cytoplasmic domain 544-976	Axitinib, Imatinib, Nilotinib, Pazopanib, Sunitinib
LCK	3932	P06239	Full-length human 1-509	
LYNa	4067	P07948	Full-length1-512	
MAP2K1	5604	Q02750	Full-length 1-393	Selumetinib, Trametinib
Met	4233	P08581	cytoplasmic domain 956-1390	Crizotinib, Cabozantinib
MPS1 (TTK)	7272	P33981	Full-length 1-857	
p38alpha**	1432	Q16539	Truncated 9-352	
PIK3CA***	5290	P42336	Full-length1-1068	
PKCeta	5583	P24723	Full-length human 1-683	
PLK1	5347	P53350	Full-length1-603	
Ret	5979	P07949	cytoplasmic domain 658-1114	Sunitinib
ROCK1	6093	Q13464	catalytic domain 1-477	
SRC	6714	P12931	Full-length 1-536	Bosutinib, Dasatinib
SYK*	6850	P43405	Full-length 1-635	
TGFBR1	7046	P36897	catalytic domain 200-503	
Tie2*	7010	Q02763	cytoplasmic domain 771-1124	
TRKB*	4915	Q16620	cytoplasmic domain 456-822	

*preactivated with ATP treatment, **co-expression with NP_002749.2 and NP_002410.1, ***co-expression with NP_852664.1

Table S2. Related to Figures 1, S2 and Supplementary Spreadsheet: Assay conditions for ePCA and kPCA

	Probe	Probe conc. [nM]	Kinase conc. [nM]	Kinetic intervals [s]*
ABL	Tracer 178	10	0.1	10
AKT1	Tracer 236	50	0.5	10
ALK	Tracer 236	10	0.5	10
AURA	Tracer 236	12.5	0.3	10
AXL	Tracer 236	25	0.2	10
BRAF	Tracer 178	125 (kPCA), 50 (ePCA)	2	10
BTK	Tracer 178	12.5	0.2	10
CDK2	Tracer 236	200	2.0	10
CDK9	Tracer 236	200 (kPCA), 50 (ePCA)	1	10
CHK1	Tracer 236	400	2.0	10
EGFR	Tracer 199	25	0.5	10
FAK	Tracer 236	25	0.3	10
FGFR1	Tracer 236	50	0.3	10
FGFR3	Tracer 236	50	0.3	10
FGFR4	Tracer 236	120	0.4	10
FLT3	Tracer 236	50	0.3	10
FMS	Tracer 236	100	0.5	10
Fyn	Tracer 236	50	0.2	10
IGF1R	Tracer 236	100	1.0	10
IKK-α	Tracer 236	70	2.0	10
IKK-β	Tracer 236	100	0.5	10
ITK	Tracer 236	50	1.0	10
JAK2	Tracer 236	5	0.4	10
KDR	Tracer 236	50	0.3	10
KIT	Tracer 222	12.5	0.3	10
LCK	Tracer 236	50	0.3	10
LYN A	Tracer 236	25	0.3	10
MAP2K1	Tracer 236	25	0.1	10
MET	Tracer 236	50	0.2	10
MPS1	Tracer 236	250 (kPCA), 50 (ePCA)	1	10
p38a	Tracer 199	50	0.124	10
PIK3CA	Tracer 314	5	0.121	10
PKCη	Tracer 236	100	0.4	120
PLK1	Tracer 236	50	4.0	10
RET	Tracer 236	25	0.3	10
ROCK1	Tracer 236	200	1.0	10
SRC	Tracer 236	25	0.3	10
Syk	Tracer 236	5	0.2	10
TGF-β1	Tracer 178	25	0.3	10
TIE2	Tracer 236	50	0.3	10
TRKB	Tracer 236	25	0.2	10

* Further reader settings: excitation - 337 nm / emission - A: 665 nm, B: 620 nm / 40 cycles

Table S3. Related to Figures 1 and S2: Binding kinetic and affinity parameters of kPCA probes

	$k_{on} [M^{-1}s^{-1}]$		$k_{off} [s^{-1}]$		$K_{D\ kin} [M]$	
	mean	error	mean	error	mean	error
ABL	$2.19 \cdot 10^6$	$1.95 \cdot 10^4$	$1.70 \cdot 10^{-4}$	$4.00 \cdot 10^{-5}$	$8.00 \cdot 10^{-11}$	$2.00 \cdot 10^{-11}$
AKT1	$8.09 \cdot 10^4$	$1.15 \cdot 10^3$	$3.99 \cdot 10^{-3}$	$9.00 \cdot 10^{-5}$	$4.93 \cdot 10^{-8}$	$1.18 \cdot 10^{-9}$
ALK	$3.75 \cdot 10^6$	$2.18 \cdot 10^5$	$9.02 \cdot 10^{-2}$	$5.33 \cdot 10^{-3}$	$2.41 \cdot 10^{-8}$	$2.30 \cdot 10^{-10}$
AURA	$1.34 \cdot 10^6$	$1.50 \cdot 10^4$	$1.55 \cdot 10^{-3}$	$6.00 \cdot 10^{-5}$	$1.16 \cdot 10^{-9}$	$4.00 \cdot 10^{-11}$
AXL	$3.31 \cdot 10^6$	$1.59 \cdot 10^5$	$1.17 \cdot 10^{-2}$	$8.10 \cdot 10^{-4}$	$3.52 \cdot 10^{-9}$	$1.30 \cdot 10^{-10}$
BRAF	$1.79 \cdot 10^6$	$3.04 \cdot 10^5$	$2.11 \cdot 10^{-1}$	$3.58 \cdot 10^{-2}$	$1.18 \cdot 10^{-7}$	$1.93 \cdot 10^{-9}$
BTk	$2.56 \cdot 10^6$	$5.20 \cdot 10^4$	$1.67 \cdot 10^{-3}$	$1.60 \cdot 10^{-4}$	$6.10 \cdot 10^{-10}$	$8.00 \cdot 10^{-11}$
CDK2	$7.94 \cdot 10^4$	$4.46 \cdot 10^3$	$1.35 \cdot 10^{-2}$	$8.70 \cdot 10^{-4}$	$1.70 \cdot 10^{-7}$	$5.85 \cdot 10^{-9}$
CDK9	$1.67 \cdot 10^5$	$4.30 \cdot 10^3$	$1.17 \cdot 10^{-2}$	$2.27 \cdot 10^{-4}$	$7.01 \cdot 10^{-8}$	$1.85 \cdot 10^{-9}$
CHK1	$2.75 \cdot 10^4$	$1.34 \cdot 10^3$	$4.53 \cdot 10^{-3}$	$3.40 \cdot 10^{-4}$	$1.65 \cdot 10^{-7}$	$1.19 \cdot 10^{-8}$
EGFR	$8.90 \cdot 10^5$	$5.12 \cdot 10^4$	$2.25 \cdot 10^{-2}$	$1.40 \cdot 10^{-3}$	$2.53 \cdot 10^{-8}$	$5.80 \cdot 10^{-10}$
FAK	$1.05 \cdot 10^6$	$5.32 \cdot 10^4$	$4.27 \cdot 10^{-2}$	$2.66 \cdot 10^{-3}$	$4.06 \cdot 10^{-8}$	$1.45 \cdot 10^{-9}$
FGFR1	$9.05 \cdot 10^5$	$9.88 \cdot 10^4$	$7.24 \cdot 10^{-2}$	$8.81 \cdot 10^{-3}$	$8.00 \cdot 10^{-8}$	$3.78 \cdot 10^{-9}$
FGFR3	$5.97 \cdot 10^5$	$3.95 \cdot 10^4$	$5.37 \cdot 10^{-2}$	$4.15 \cdot 10^{-3}$	$8.99 \cdot 10^{-8}$	$3.56 \cdot 10^{-9}$
FGFR4	$2.29 \cdot 10^5$	$2.29 \cdot 10^4$	$1.69 \cdot 10^{-1}$	$1.67 \cdot 10^{-2}$	$7.37 \cdot 10^{-7}$	$2.20 \cdot 10^{-8}$
FLT3	$6.89 \cdot 10^4$	$6.84 \cdot 10^2$	$4.23 \cdot 10^{-3}$	$7.00 \cdot 10^{-5}$	$6.15 \cdot 10^{-8}$	$9.10 \cdot 10^{-10}$
FMS	$1.10 \cdot 10^5$	$3.30 \cdot 10^3$	$9.21 \cdot 10^{-3}$	$3.40 \cdot 10^{-4}$	$8.38 \cdot 10^{-8}$	$1.94 \cdot 10^{-9}$
Fyn	$3.00 \cdot 10^6$	$6.03 \cdot 10^5$	$1.43 \cdot 10^{-1}$	$3.00 \cdot 10^{-2}$	$4.75 \cdot 10^{-8}$	$1.92 \cdot 10^{-9}$
IGF1R	$4.76 \cdot 10^5$	$5.77 \cdot 10^4$	$1.32 \cdot 10^{-1}$	$1.45 \cdot 10^{-2}$	$2.78 \cdot 10^{-7}$	$1.95 \cdot 10^{-8}$
IKK- α	$1.65 \cdot 10^5$	$2.91 \cdot 10^3$	$1.48 \cdot 10^{-2}$	$3.10 \cdot 10^{-4}$	$9.01 \cdot 10^{-8}$	$1.11 \cdot 10^{-9}$
IKK- β	$3.16 \cdot 10^5$	$3.29 \cdot 10^4$	$4.89 \cdot 10^{-2}$	$5.13 \cdot 10^{-3}$	$1.54 \cdot 10^{-7}$	$7.85 \cdot 10^{-9}$
ITK	$1.01 \cdot 10^6$	$8.06 \cdot 10^4$	$7.77 \cdot 10^{-2}$	$6.80 \cdot 10^{-3}$	$7.68 \cdot 10^{-8}$	$3.30 \cdot 10^{-9}$
JAK2	$3.43 \cdot 10^6$	$7.05 \cdot 10^4$	$1.36 \cdot 10^{-2}$	$3.20 \cdot 10^{-4}$	$3.96 \cdot 10^{-9}$	$6.00 \cdot 10^{-11}$
KDR	$1.44 \cdot 10^5$	$1.68 \cdot 10^3$	$4.46 \cdot 10^{-3}$	$9.00 \cdot 10^{-5}$	$3.09 \cdot 10^{-8}$	$4.40 \cdot 10^{-10}$
KIT	$2.42 \cdot 10^6$	$1.19 \cdot 10^5$	$2.00 \cdot 10^{-2}$	$1.12 \cdot 10^{-3}$	$8.28 \cdot 10^{-9}$	$2.30 \cdot 10^{-10}$
LCK	$1.58 \cdot 10^6$	$2.27 \cdot 10^5$	$1.22 \cdot 10^{-1}$	$1.86 \cdot 10^{-2}$	$7.73 \cdot 10^{-8}$	$3.58 \cdot 10^{-9}$
LYN A	$2.27 \cdot 10^6$	$3.42 \cdot 10^5$	$8.54 \cdot 10^{-2}$	$1.36 \cdot 10^{-2}$	$3.76 \cdot 10^{-8}$	$1.63 \cdot 10^{-9}$
MAP2K1	$1.71 \cdot 10^5$	$1.39 \cdot 10^3$	$2.10 \cdot 10^{-4}$	$4.00 \cdot 10^{-5}$	$1.23 \cdot 10^{-9}$	$2.30 \cdot 10^{-10}$
MET	$2.57 \cdot 10^5$	$8.43 \cdot 10^3$	$2.87 \cdot 10^{-2}$	$1.00 \cdot 10^{-3}$	$1.12 \cdot 10^{-7}$	$1.57 \cdot 10^{-9}$
MPS1	$1.16 \cdot 10^6$	$1.24 \cdot 10^5$	$1.48 \cdot 10^{-1}$	$1.58 \cdot 10^{-2}$	$1.28 \cdot 10^{-7}$	$2.79 \cdot 10^{-9}$
p38a	$2.17 \cdot 10^6$	$2.63 \cdot 10^5$	$8.99 \cdot 10^{-2}$	$1.11 \cdot 10^{-2}$	$4.14 \cdot 10^{-8}$	$7.80 \cdot 10^{-10}$
PIK3CA	$8.62 \cdot 10^6$	$1.74 \cdot 10^5$	$4.98 \cdot 10^{-3}$	$1.60 \cdot 10^{-4}$	$5.80 \cdot 10^{-10}$	$1.00 \cdot 10^{-11}$
PKC η	$4.64 \cdot 10^3$	$8.77 \cdot 10^1$	$3.04 \cdot 10^{-4}$	$1.00 \cdot 10^{-5}$	$6.55 \cdot 10^{-8}$	$1.89 \cdot 10^{-9}$
PLK1	$3.03 \cdot 10^6$	$2.75 \cdot 10^5$	$1.95 \cdot 10^{-2}$	$1.94 \cdot 10^{-3}$	$6.44 \cdot 10^{-9}$	$2.30 \cdot 10^{-10}$
RET	$9.78 \cdot 10^5$	$2.88 \cdot 10^4$	$7.40 \cdot 10^{-3}$	$2.90 \cdot 10^{-4}$	$7.57 \cdot 10^{-9}$	$1.70 \cdot 10^{-10}$
ROCK1	$9.21 \cdot 10^4$	$3.42 \cdot 10^3$	$3.26 \cdot 10^{-2}$	$1.21 \cdot 10^{-3}$	$3.54 \cdot 10^{-7}$	$8.44 \cdot 10^{-9}$
SRC	$6.01 \cdot 10^6$	$2.26 \cdot 10^6$	$1.69 \cdot 10^{-1}$	$6.48 \cdot 10^{-2}$	$2.81 \cdot 10^{-8}$	$1.42 \cdot 10^{-9}$
Syk	$2.73 \cdot 10^6$	$5.13 \cdot 10^4$	$3.83 \cdot 10^{-3}$	$1.30 \cdot 10^{-4}$	$1.40 \cdot 10^{-9}$	$4.00 \cdot 10^{-11}$
TGF- β 1	$2.10 \cdot 10^6$	$2.29 \cdot 10^5$	$8.55 \cdot 10^{-2}$	$9.44 \cdot 10^{-3}$	$4.06E \cdot 10^{-8}$	$7.10 \cdot 10^{-10}$
TIE2	$2.43 \cdot 10^6$	$1.01 \cdot 10^6$	$2.72 \cdot 10^{-1}$	$1.15 \cdot 10^{-1}$	$1.12 \cdot 10^{-7}$	$6.21 \cdot 10^{-9}$
TRKB	$7.68 \cdot 10^6$	$1.32 \cdot 10^4$	$1.17 \cdot 10^{-3}$	$9.00 \cdot 10^{-5}$	$1.52 \cdot 10^{-9}$	$1.10 \cdot 10^{-10}$

Table S4. Related to all Figures and Tables: Compounds screened in this study

Compound name	CAS Number	Compound name	CAS Number
3-Methyladenine	5142-23-4	Bosutinib (SKI-606)	380843-75-4
A66	1166227-08-2	Brivanib (BMS-540215)	649735-46-6
A-674563	552325-73-2	Brivanib alaninate (BMS-582664)	649735-63-7
A-769662	844499-71-4	BS-181 HCl	1397219-81-6
AEE788 (NVP-AEE788)	497839-62-0	BX-795	702675-74-9
Afatinib (BIBW2992)	439081-18-2	BX-912	702674-56-4
AG-1024	65678-07-1	BYL719	1217486-61-7
AG-1478 (Tyrphostin AG-1478)	153436-53-4	CAL-101 (GS-1101)	870281-82-6
AG-490	133550-30-8	CAY10505	1218777-13-9
AMG 900	945595-80-2	CCT128930	885499-61-6
AMG-208	1002304-34-8	CCT129202	942947-93-5
AMG458	913376-83-7	CCT137690	1095382-05-0
Amuvatinib (MP-470)	850879-09-3	Cediranib (AZD2171)	288383-20-0
Apatinib (YN968D1)	811803-05-1	CEP33779	1257704-57-6
ARQ 197 (Tivantinib)	905854-02-6	CH5424802	1256580-46-7
ARRY334543	845272-21-1	CHIR-124	405168-58-3
Arry-380	937265-83-3	CHIR-98014	252935-94-7
AS-252424	900515-16-4	CI-1033 (Canertinib)	267243-28-7
AS-604850	648449-76-7	CI-1040 (PD184352)	212631-79-3
AS-605240	648450-29-7	CP 673451	343787-29-1
AS703026 (pimasertib)	1236699-92-5	CP-724714	537705-08-1
AST-1306	1050500-29-2	Crenolanib (CP-868596)	670220-88-9
AT7519	844442-38-2	Crizotinib (PF-02341066)	877399-52-5
AT7867	857531-00-1	CX-4945 (Silmisertib)	1009820-21-6
AT9283	896466-04-9	CYC116	693228-63-6
Aurora A Inhibitor I	1158838-45-9	Cyt387	1056634-68-4
Axitinib	319460-85-0	Dabrafenib (GSK2118436)	1195765-45-7
AZ 960	905586-69-8	Dacomitinib (PF299804 - PF-00299804)	1110813-31-4
AZ628	878739-06-1	Danuserib (PHA-739358)	827318-97-8
AZD2014	1009298-59-2	Dasatinib (BMS-354825)	302962-49-8
AZD4547	1035270-39-3	DCC-2036 (Rebastinib)	1020172-07-9
AZD5438	602306-29-6	Deforolimus (Ridaforolimus)	572924-54-0
AZD6244 (Selumetinib)	606143-52-6	Desmethyl Erlotinib (CP-473420)	183320-51-6
AZD7762	860352-01-8	Dinaciclib (SCH727965)	779353-01-4
AZD8055	1009298-09-2	Dovitinib (TKI-258)	405169-16-6
AZD8330	869357-68-6	Dovitinib Dilactic acid (TKI258 Dilactic acid)	852433-84-2
AZD8931	848942-61-0	E7080 (Lenvatinib)	417716-92-8
Barasertib (AZD1152-HQPA)	722544-51-6	ENMD-2076	1291074-87-7
Baricitinib (LY3009104 - incb28050)	1187594-09-7	Enzastaurin (LY317615)	170364-57-5
BEZ235 (NVP-BEZ235)	915019-65-7	Erlotinib HCl	183319-69-9
BGJ398 (NVP-BGJ398)	872511-34-7	Everolimus (RAD001)	159351-69-6
BI 2536	755038-02-9	Flavopiridol HCl	131740-09-5
BI6727 (Volasertib)	755038-65-4	Foretinib (GSK1363089 - XL880)	849217-64-7
BIBF1120 (Vargatef)	656247-17-5	GDC-0068	1001264-89-6
BIRB 796 (Doramapimod)	285983-48-4	GDC-0879	905281-76-7
BIX 02188	1094614-84-2	GDC-0941	957054-30-7
BIX 02189	1094614-85-3	GDC-0980 (RG7422)	1032754-93-0
BKM120 (NVP-BKM120)	944396-07-0	Gefitinib (Iressa)	184475-35-2
BMS 777607	1025720-94-8	Golvatinib (E7050)	928037-13-2
BMS 794833	1174046-72-0	GSK1059615	958852-01-2
BMS-265246	582315-72-8	GSK1070916	942918-07-2
BMS-599626 (AC480)	714971-09-2		

Compound name	CAS Number
GSK1120212 (Trametinib)	871700-17-3
GSK1838705A	1116235-97-2
GSK1904529A	1089283-49-7
GSK2126458	1086062-66-9
GSK461364	929095-18-1
GSK690693	937174-76-0
Hesperadin	422513-13-1
HMN-214	173529-46-9
IC-87114	371242-69-2
Imatinib (Gleevec)	152459-95-5
Imatinib Mesylate	220127-57-1
IMD 0354	978-62-1
INCB28060	1029712-80-8
Indirubin	479-41-4
INK 128 (MLN0128)	1224844-38-5
JNJ-38877605	943540-75-8
JNJ-7706621	443797-96-4
Ki8751	228559-41-9
KRN 633	286370-15-8
Ku-0063794	938440-64-3
KU-55933	587871-26-9
KU-60019	925701-49-1
KW 2449	1000669-72-6
KX2-391	897016-82-9
Lapatinib Ditosylate (Tykerb)	388082-77-7
LDN193189	1062368-24-4
Linifanib (ABT-869)	796967-16-3
Linsitinib (OSI-906)	867160-71-2
LY2228820	862507-23-1
LY2603618 (IC-83)	911222-45-2
LY2784544	1229236-86-5
LY294002	154447-36-6
Masitinib (AB1010)	790299-79-5
MGCD-265	875337-44-3
Milciclib (PHA-848125)	802539-81-7
MK-2206 2HCl	1032350-13-2
MK-2461	917879-39-1
MK-5108 (VX-689)	1010085-13-8
MLN8054	869363-13-3
MLN8237 (Alisertib)	1028486-01-2
Motesanib Diphosphate (AMG-706)	857876-30-3
Mubritinib (TAK 165)	366017-09-6
Neratinib (HKI-272)	698387-09-6
Nilotinib (AMN-107)	641571-10-0
NU7441 (KU-57788)	503468-95-9
NVP-ADW742	475488-23-4
NVP-BGT226	1245537-68-1
NVP-BHG712	940310-85-0
NVP-BSK805	1092499-93-8
NVP-BVU972	1185763-69-2
NVP-TAE226	761437-28-9
ON-01910	1225497-78-8
OSI-027	936890-98-1
OSI-420	183320-51-6
OSI-930	728033-96-3

Compound name	CAS Number
OSU-03012	742112-33-0
Palomid 529	914913-88-5
Pazopanib HCl	635702-64-6
PCI-32765 (Ibrutinib)	936563-96-1
PD 0332991 (Palbociclib) HCl	827022-32-2
PD0325901	391210-10-9
PD153035 HCl	183322-45-4
PD173074	219580-11-7
PD318088	391210-00-7
PD98059	167869-21-8
Pelitinib (EKB-569)	257933-82-7
PF-00562271	939791-38-5
PF-03814735	942487-16-3
PF-04217903	956905-27-4
PF-04691502	1013101-36-4
PF-05212384 (PKI-587)	1197160-78-3
PH-797804	586379-66-0
PHA-665752	477575-56-7
PHA-680632	398493-79-3
PHA-767491	845714-00-3
PHA-793887	718630-59-2
Phenformin HCl	834-28-6
PHT-427	1191951-57-1
PI-103	371935-74-9
Piceatannol	10083-24-6
PIK-293	900185-01-5
PIK-294	900185-02-6
PIK-75	372196-77-5
PIK-90	677338-12-4
PIK-93	593960-11-3
PKI-402	1173204-81-3
PLX-4720	918505-84-7
Ponatinib (AP24534)	943319-70-8
PP-121	1092788-83-4
PP242	1092351-67-1
Quercetin (Sophoretin)	117-39-5
Quizartinib (AC220)	950769-58-1
R406	841290-81-1
R406 (free base)	841290-80-0
R788 (Fostamatinib)	901119-35-5
R935788 (Fostamatinib disodium - R788 disodium)	1025687-58-4
Raf265 derivative	n.a.; Selleckchem Cat.No. S2200
Rapamycin (Sirolimus)	53123-88-9
Roscovitine (Seliciclib - CYC202)	186692-46-6
Ruxolitinib (INCB018424)	941678-49-5
SAR131675	1433953-83-3
Saracatinib (AZD0530)	379231-04-6
SB 202190	152121-30-7
SB 203580	152121-47-6
SB 216763	280744-09-4
SB 415286	264218-23-7
SB 431542	301836-41-9
SB 525334	356559-20-1

Compound name	CAS Number
SB590885	405554-55-4
Semaxanib (SU5416)	194413-58-6
SGX-523	1022150-57-7
SNS-032 (BMS-387032)	345627-80-7
SNS-314	1146618-41-8
Sotrastaurin (AEB071)	425637-18-9
SP600125	129-56-6
Staurosporine	62996-74-1
SU11274	658084-23-2
Sunitinib Malate (Sutent)	341031-54-7
TAE684 (NVP-TAE684)	761439-42-3
TAK-285	871026-44-7
TAK-733	1035555-63-5
TAK-901	934541-31-8
Tandutinib (MLN518)	387867-13-2
Temsirolimus (Torisel)	162635-04-3
TG 100713	925705-73-3
TG100-115	677297-51-7
TG101209	936091-14-4
TG101348 (SAR302503)	936091-26-8
TGX-221	663619-89-4
Thiazovivin	1226056-71-8
Tideglusib	865854-05-3
Tie2 kinase inhibitor	948557-43-5
Tivozanib (AV-951)	475108-18-0
Tofacitinib (CP-690550 - Tasocitinib)	477600-75-2
Tofacitinib citrate (CP-690550 citrate)	540737-29-9
Torin 1	1222998-36-8
Torin 2	1223001-51-1
TPCA-1	507475-17-4
Triciribine (Triciribine phosphate)	35943-35-2
TSU-68 (SU6668)	252916-29-3
TWS119	601514-19-6
Tyrphostin AG 879 (AG 879)	148741-30-4
U0126-EtOH	1173097-76-1
Vandetanib (Zactima)	443913-73-3
Vatalanib 2HCl (PTK787)	212141-51-0
Vemurafenib (PLX4032)	918504-65-1
VX-680 (MK-0457 - Tozasertib)	639089-54-6
VX-702	745833-23-2
WAY-600	1062159-35-6
WHI-P154	211555-04-3
Wortmannin	19545-26-7
WP1066	857064-38-1
WP1130	856243-80-6
WYE-125132	1144068-46-1
WYE-354	1062169-56-5
WYE-687	1062161-90-3
WZ3146	1214265-56-1
WZ4002	1213269-23-8
WZ8040	1214265-57-2
XL147	956958-53-5
XL-184 (Cabozantinib)	849217-68-1
XL765	1349796-36-6

Compound name	CAS Number
Y-27632 2HCl	129830-38-2
YM201636	371942-69-7
ZM 336372	208260-29-1
ZM-447439	331771-20-1
ZSTK474	475110-96-4

Table S5. Related to Figures 3 and S4: Area under the curve (AUC) and maximal target occupancy (TO_{max}) values for the simulated TO dynamics traces shown in Fig. 3c and Fig. S4d.

Drug	Administration schedule	Kinase	AUC of TO trace simulated with PK-K _D model	AUC of TO trace simulated with PK-BK model	Maximal TO for simulation with PK-K _D model	Maximal TO for simulation with PK-BK model
Dasatinib	1 mg/kg / 12h (day 3)	ABL	11.99	11.99	0.9997	0.9997
		FYN	11.55	11.56	0.9880	0.9878
		KIT	9.58	9.63	0.9237	0.9222
		LCK	10.76	10.78	0.9650	0.9641
		SRC	11.90	11.91	0.9970	0.9974
		BTK	10.39	10.41	0.9530	0.9517
Dasatinib	1 mg/kg / 48h (day 7-8)	ABL	47.19	47.52	0.9996	0.9996
		FYN	33.65	36.72	0.9862	0.9848
		KIT	17.65	19.56	0.9126	0.9042
		LCK	24.45	27.05	0.9590	0.9554
		SRC	43.04	44.75	0.9970	0.9968
		BTK	21.82	24.11	0.9450	0.9402
'Dasatinib-like' but clearance↑ (4x inter-compartmental CL; 10x plasma clearance)	0.2 mg/kg / 24h (day 3)	ABL	10.28	20.28	0.9920	0.9830
		FYN	3.44	4.98	0.7742	0.6206
		KIT	1.00	1.14	0.3343	0.1705
		LCK	1.82	2.25	0.5320	0.3524
		SRC	6.33	11.24	0.9430	0.8881
		BTK	1.47	1.75	0.4550	0.2853
Ibrutinib (assuming	8 mg/kg / 24h	BTK	15.58	18.22	0.9953	0.9939
		LCK	5.62	5.83	0.9343	0.9322

kPCA k _{off})	(day 4)	EGFR	12.74	14.22	0.9906	0.9894
		ITK	3.34	5.54	0.8410	0.4785
Ibrutinib (assuming covalent binding)		BTK	-	24.00	-	1.000
		LCK	-	5.81	-	0.9326
		EGFR	-	24.00	-	1.000
		ITK	-	23.66	-	0.9877

Table S6. Related to Figures 5 and S6: Matched molecular pairs screened in this study

Desmethyl Erlotinib → Erlotinib	
PD153035 → Tyrphostin AG-1478	
WHI-P154 → PD153035	
BIX 02189 → BIX 02188 ($k_{on} \uparrow$, $K_D \downarrow$)	
Crenolanib → CP 673451 ($k_{on} \uparrow$, $k_{off} \uparrow$)	
WZ4002 → WZ3146 ($k_{on} \uparrow$, $K_D \downarrow$)	

Table S7. Related to Figures 5, 6, S6 and S7: Comparison of quality of GFA models with 1) molecular properties as descriptors for binding parameters considering all kinase-inhibitor pairs, kinase-inhibitor pairs with $pK_D > 8$, or individual kinases (with $pK_D > 8$), 2) molecular properties or KLIFS binding mode information and/or KLIFS interaction fingerprints as descriptors for binding parameters (considering only kinase-inhibitor pairs with PDB and KLIFS entry). r^2 represents the percentage of variability of the data that can be described by the model.

Comparison	r^2 of GFA model for	$\log(k_{on})$	$-\log(k_{off})$	$-\log(K_D)$
1	all kinase-inhibitor pairs	0.06	0.07	0.06
	all kinase- inhibitor pairs with $pK_D > 8$	0.16	0.16	0.06
	RET- inhibitor pairs with $pK_D > 8$	0.34	0.67	
	PIK3CA- inhibitor pairs with $pK_D > 8$	0.86	0.51	
	KDR- inhibitor pairs with $pK_D > 8$	0.30	0.45	
	AurA- inhibitor pairs with $pK_D > 8$	0.25	0.41	
	EGFR- inhibitor pairs with $pK_D > 8$	0.19	0.34	
	JAK2- inhibitor pairs with $pK_D > 8$	0.23	0.31	
2	Molecular properties	0.41	0.38	0.27
	KLIFS binding mode	0.43	0.33	0.15
	KLIFS interaction fingerprints	0.74	0.68	0.54
	KLIFS interaction fingerprints & molecular properties	0.79	0.69	0.58

SUPPLEMENTARY REFERENCES

- (1) Wu, P.; Nielsen, T. E.; Clausen, M. H. *Trends Pharmacol Sci* **2015**, 36, 422.