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Promiscuity of inhibitors of human protein kinases at varying data confidence levels and test frequencies

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More than 141 000 inhibitors of human kinases and their activity data were assembled to perform an in-depth analysis of inhibitor promiscuity (single- *versus* multi-kinase activity) at varying activity data confidence levels. For ~20% of these inhibitors, it was also possible to consider test frequency and inactivity information. Only small subsets of highly promiscuous inhibitors were identified. Nearly 95% of more than 45 000 inhibitors with high-confidence data were only active against one or at most two kinases. At decreasing data confidence levels, more than 92 000 kinase inhibitors were on average active against two kinases. When taking all activity information without any restrictions into account, the mean promiscuity degree of kinase inhibitors was less than four and notably biased by small numbers of highly promiscuous inhibitors. Even under these conditions, more than 70% of all inhibitors were active against a single kinase. There was only small-scale progression of inhibitor promiscuity when data confidence criteria were iteratively removed during the analysis. Furthermore, the majority of inhibitors that were tested against 10 to 20 different kinases were only active against a single kinase. The results of our activity data-driven analysis indicate that promiscuity of kinase inhibitors cannot generally be assumed. Many inhibitors retain single-kinase activity at decreasing data confidence criteria or increasing test frequency. Hence, on the basis of currently available data, many kinase inhibitors are selective, which is an important aspect for drug development.

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Introduction

Protein kinases are major drug targets^{1,2} and the promiscuity of classical ATP site-directed (type I) kinase inhibitors continues to be debated.^{3,4} In this context, promiscuity is rationalized as the ability of an inhibitor to interact with multiple kinases. Given that the ATP site is largely conserved across the kinase, it is often thought that these inhibitors might be active against many different kinases. The human kinase comprises 518 kinases,⁵ excluding complexes and isoforms. Some highly promiscuous kinase inhibitors were identified.⁶ Among these were compounds that have proven effective as anti-cancer agents for which promiscuity and ensuing polypharmacology often play a decisive role.³

On the other hand, recent global analyses of high-confidence activity data for publicly available inhibitors of the human kinase have revealed that the majority of these inhibitors were only annotated with one or two kinase targets.^{6,7} Data

incompleteness and the unavailability of test frequency information in the literature might at least in part explain these findings, but it is also conceivable that the promiscuity of many ATP site-directed kinase inhibitors is indeed lower than often thought.

Data-driven promiscuity assessment benefits from concentrating on high-confidence activity data. This requires careful data curation, but provides the best possible basis for arriving at sound promiscuity estimates. Although such estimates are intrinsically conservative, they are least influenced by experimental heterogeneity. For compound data mining, this is an important aspect to consider.⁶

Of course, further differentiated data mining strategies might be considered. For example, given that the majority of kinase inhibitors were not found to be promiscuous on the basis of high-confidence activity data, one might assess the issue of data sparseness by softening confidence criteria and taking increasing amounts of activity data into account. This would make it possible to evaluate an anticipated progression of inhibitor promiscuity as activity data increases and determine its magnitude. Thus, monitoring promiscuity progression was a primary goal of our current analysis.

In addition, our study was further motivated by the remarkable increase in the number of kinase inhibitors that are becoming publicly available.⁸ For example, our survey of

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inhibitors of the human kinome in 2015⁶ was based upon nearly 19 000 kinase inhibitors for which high-confidence activity data were available in ChEMBL,⁹ the major public repository of compounds from the medicinal chemistry literature. These inhibitors were active against a total of 266 human kinases.⁶ However, early in 2017, more than 45 000 kinase inhibitors with high-confidence activity data were available in ChEMBL.⁸ These inhibitors were active against 286 human kinases. Hence, within merely two years, the number of public kinase inhibitors more than doubled, although kinome coverage only slightly increased. Notably, 70% of the qualifying inhibitors available in 2015 were only annotated with a single kinase.⁶ For the much larger number of inhibitors available in early 2017, this proportion further increased to 76%.⁸

Moreover, we have also addressed the issue of promiscuity *versus* test frequency by including compounds from PubChem BioAssays¹¹ in our analysis. Extensively tested screening compounds with activity in kinase assays were identified and their target-based assay frequency and promiscuity were determined. Thus, in addition to studying promiscuity at varying data confidence levels for large numbers of kinase inhibitors, it was also possible to compare inhibitor promiscuity in the presence and absence of test frequency information. The results of our analysis are reported in the following.

Methods and materials

Human kinase targets

Human kinases were extracted from UniProt¹⁰ that organizes both human and mouse kinases and contains 504 human entries. UniProt IDs for human kinases were linked to ChEMBL⁹

and PubChem¹¹ target IDs to consistently associate compounds with kinases.

Kinase inhibitors

Classical (type I) kinase inhibitors are ATP site-directed, compete with ATP, and bind to the “DFG-in” conformation of the activation loop in the active site.¹² Although other classes (type II–IV) of kinase inhibitors exist,^{12,13} which partly bind to regions distant from the ATP site and are allosteric in nature (type III–IV),^{12–14} the vast majority of publicly available kinase inhibitors, estimated to be more than 98%,⁶ are type I compounds. Thus, for large-scale activity data mining, other types of inhibitors can currently be neglected.

Compounds active against at least one of all available human kinases were selected from ChEMBL release 22 and a subset of screening compounds from PubChem BioAssays.¹¹ This subset of PubChem consisted of 437 257 compounds that were tested in both primary assays (percentage of inhibition from a single dose) and confirmatory assays (dose-response assays yielding IC₅₀ values).¹⁶ From this subset of extensively assayed compounds, with a mean and median of 411 and 437 assays per compound, respectively,¹⁵ kinase inhibitors were selected. Kinase inhibitors from ChEMBL were analyzed under varying activity data confidence criteria, as specified below. From ChEMBL, inactivity records from assays or test frequency data cannot be obtained. However, for PubChem inhibitors, the number of kinases against which they were tested (also referred to as test frequency) was determined and taken into account in assessing their promiscuity. Fig. 1 summarizes the target and compound selection process.

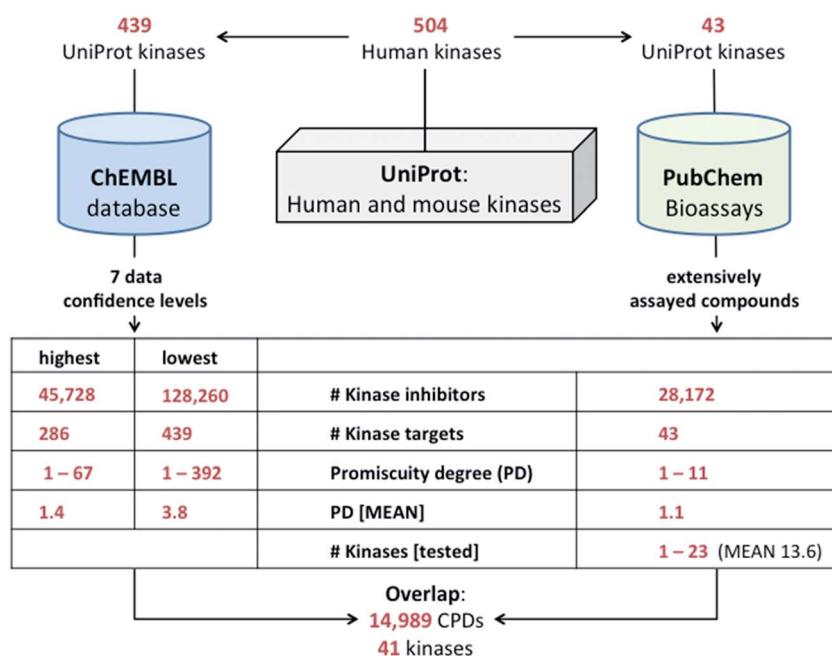


Fig. 1 Kinases and inhibitors. Human kinases were mapped and inhibitors extracted from ChEMBL and PubChem. On the left, kinase inhibitor data obtained from ChEMBL are summarized for varying selection criteria (according to Fig. 2). On the right, extensively tested inhibitors of kinases with ChEMBL target IDs available in PubChem are reported.

Data confidence criteria

From ChEMBL, inhibitors can be selected following a hierarchy of seven database-specific criteria⁹ to gradually increase (or decrease) the degree of activity data confidence,¹⁶ as illustrated in Fig. 2. For our analysis, we first extracted kinase inhibitors on the basis of high activity data confidence, corresponding to confidence level 1 in Fig. 2. Then, selection criteria were iteratively removed (from the top to bottom in Fig. 2) to gradually transition from high- to low-confidence data, producing a total of seven different confidence levels (1–7). The following criteria were applied:

- (i) Direct interaction assays with highest confidence: assay relationship type 'D', assay confidence score '9';
- (ii) Specific targets: target type 'SINGLE PROTEIN';
- (iii) Defined activity measurements: activity type ' K_i ' or ' IC_{50} ';
- (iv) Specified activity values: standard relation '=';
- (v) Standard activity unit: 'nM';
- (vi) Activity comments: removal of compounds designated as inconclusive, not active, inactive, not evaluated/determined.
- (vii) Kinase organism annotation: '*Homo sapiens*'.

Accordingly, kinase inhibitors at the highest confidence level 1 were required to meet all seven selection criteria, yielding the smallest set. By contrast, for kinase inhibitors at the lowest confidence level 7, all available activity data were taken into account, without any confidence measures, hence producing the largest set of inhibitors. The increase in the number of inhibitors between confidence level 1 and 7 was not dependent on the order in which selection criteria were applied. Each confidence level defines activity criteria for ChEMBL compounds. We note that K_i and IC_{50} values were not separately considered here to support increasing promiscuity levels for inhibitors.

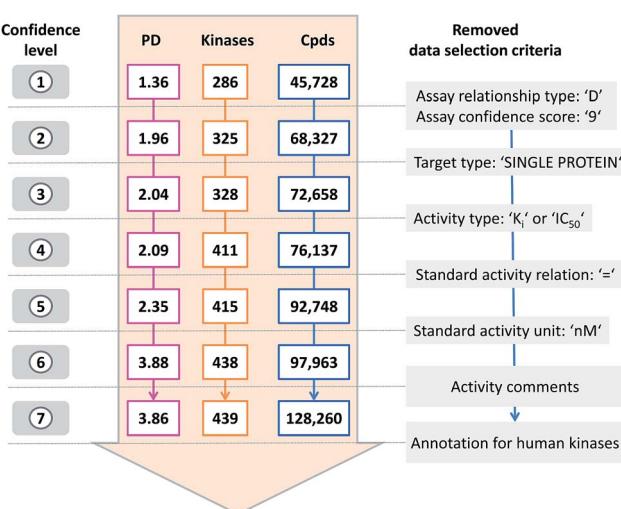


Fig. 2 Data selection criteria and confidence levels. A sequence of ChEMBL selection criteria (right; detailed in the text) is applied to obtain kinase inhibitors at varying activity data confidence levels (left). From the top to the bottom (i.e., level 1 to 7), data confidence decreases and the number of qualifying inhibitors increases. For each confidence level, kinase and compound statistics and the mean promiscuity degree (PD) of the inhibitors are reported.

In PubChem, similar data confidence criteria cannot be applied. However, in addition to focusing on extensively assayed inhibitors, the requirement of qualifying PubChem compounds to be tested in both primary and confirmatory assays also represented a data confidence criterion. For example, under these conditions, low-confidence kinase profiling data from single experiments incorporated into PubChem did not qualify for the analysis. For kinase inhibitors from PubChem, activity annotations from primary and confirmatory assays against different kinase targets were combined to yield upper level promiscuity estimates.

Promiscuity degree

For our analysis, the promiscuity degree (PD) of an inhibitor was defined as the number of human kinases it was active against.

Data mining and analysis

All calculations were performed using in-house scripts and KNIME¹⁷ protocols with the aid of the OpenEye¹⁸ chemistry toolkit.

Results and discussion

Analysis scheme

Our comprehensive analysis of kinase inhibitors and their activity data in ChEMBL and PubChem is summarized in Fig. 1. We first mapped UniProt IDs of human kinases to ChEMBL and PubChem. A total of 439 kinase entries were detected in ChEMBL. In addition, the extensively assayed subset of PubChem included compounds tested against 43 kinases. For all human kinases in ChEMBL and PubChem, inhibitors were systematically identified and assigned to UniProt IDs. Then their activity records were analyzed.

ChEMBL

We first analyzed data from ChEMBL. Compounds and activity data in ChEMBL primarily originate from the medicinal chemistry literature and are manually curated. Given the source of the data, no records of test frequency or inactivity are provided in this database.

Compound and kinase statistics. Considering all activity data available in ChEMBL for human kinases, corresponding to the lowest data confidence level 7 in Fig. 2, we identified a total of 128 260 inhibitors for 439 kinases. For 45 728 of these inhibitors, which were active against 286 human kinases, high-confidence activity data were also available, corresponding to confidence level 1 in Fig. 2. Hence, an unprecedentedly large number of inhibitors was analyzed for nearly 300 (level 1) and more than 400 (level 7) human kinases.

Data confidence levels. Inhibitors at confidence level 1 and level 7 were active against one to 67 and 392 kinases, respectively, and thus included at least some highly promiscuous compounds. However, the mean PD of inhibitors at level 1 was only 1.4 and moderately increased to 3.9 at level 7. At each confidence level, activity annotations change and under

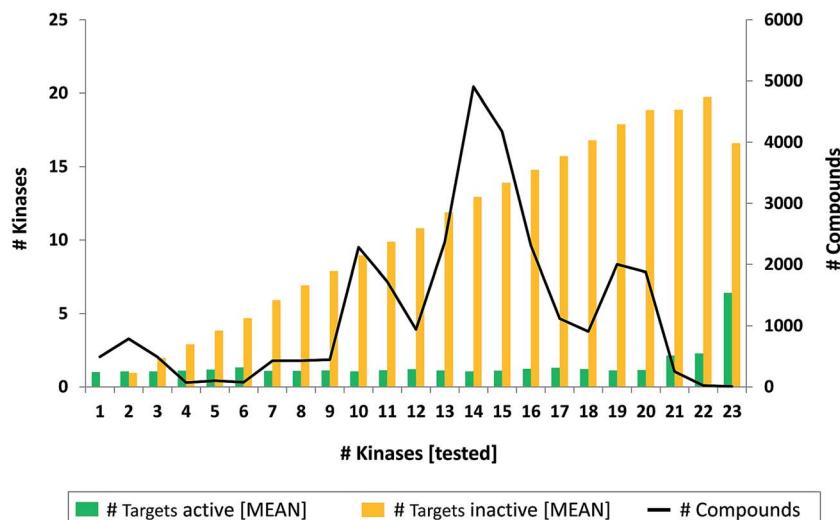


Fig. 3 Promiscuity of kinase inhibitors from PubChem. The promiscuity of inhibitors tested against increasing numbers (1–23) of kinases is reported. For each number of tested kinases, the mean number of kinases against which the inhibitors were active (green) or inactive (yellow) is given (left axis). The black line shows the distribution of compounds (right axis) over different assays. For example, about 5000 inhibitors were tested in 14 different kinase assays.

decreasing data confidence, more compounds should be annotated as active. Fig. 2 shows how the number of qualifying inhibitors and their kinase coverage gradually increased from level 1 to 7 when data confidence criteria were iteratively removed. It also reports the increase in mean PD values under decreasing activity data confidence. Interestingly, the mean PD only slightly increased from 1.4 to 2.4 over confidence levels 1–5, which accounted for assay and activity measurement confidence. Thus, as long as assay or measurement criteria were specified, inhibitors were on average only annotated with one or two kinases, which applied to a large number of 92 748 inhibitors. An increase in the mean PD from 2.4 to 3.9 was only observed when standard activity units were no longer required and all types of measurements were considered including, for example, percentage of inhibition or residual activity. Thus, as long as at least the standard activity unit (nM) was reported in activity records, the mean promiscuity of human kinase inhibitors from ChEMBL was low, even if no additional data confidence criteria were applied. At highest measurement confidence, corresponding to confidence level 3, the mean PD value was 2.0 and decreased to 1.4 when highest assay confidence was also required (proceeding from level 3 to 1). For all three mean promiscuity values of 1.4 (level 1), 2.0 (3), and 3.9 (7), the corresponding median values were 1.0, hence indicating that small numbers of highly promiscuous inhibitors were mostly responsible for the PD increase, especially from 2.0 to 3.9.

Distribution of promiscuity degrees. The proportions of inhibitors with different PD values over all confidence levels was determined. As discussed above, the number of qualifying inhibitors substantially increased from level 1 to 7. However, the proportion of inhibitors with single-kinase activity (PD 1) remained remarkably constant at the 70% level. Hence, even at lowest data confidence, the majority of inhibitors were only

annotated with a single kinase. Moreover, an additional 20% to less than 30% of the inhibitors fell into the PD interval 2–4 across all confidence levels. By contrast, inhibitors with activity against at least five kinases were generally rare. At confidence level 1, 533 of the inhibitors (1.2%) were active against at least five kinases and 101 (0.2%) against 10 or more kinases. At confidence level 7, 7911 (6.2%) and 4393 (3.4%) of the corresponding 128 260 inhibitors were active against at least five and at least 10 kinases, respectively. At decreasing levels of data confidence, only small subsets of highly promiscuous inhibitors were detected. Thus, the increase in mean PD values from 2.4 to 3.9 observed at level 5 and 7 was largely due to small numbers of highly promiscuous inhibitors (which may also include activity artifacts), as indicated by the difference between increasing mean and constant median PD values discussed above.

PubChem

Given that no records of assay frequency or inactivity are available in ChEMBL, we extended the analysis to the PubChem BioAssay database.¹¹ Inactivity or assay frequency data for screening compounds are not provided in PubChem records either, but can be obtained by determining in which assays individual compounds have been tested and found to be active. We analyzed 437 257 PubChem compounds that were extensively tested in both primary and confirmatory assays, identified inhibitors of human kinases, and determined their target-based test frequency.

Compounds and kinases. As reported in Fig. 1, the set of extensively assayed compounds included 28 172 kinase inhibitors with activity against 43 different kinases. These inhibitors from PubChem were tested against one to 23 human kinases, with a mean of 13.6 kinases per compound. Notably, 14 989 of the inhibitors detected in PubChem, with activity against 41 different kinases, were also found in ChEMBL (owing to the fact



ChEMBL compounds			
Confidence level			
1 – PD	5	1	1
7 – PD	7	1	1
1 – PD	1	1	1
7 – PD	3	1	1
PubChem compounds			
# Kinases			
active	3	1	1
inactive	10	9	17

Fig. 4 Kinase inhibitors with varying promiscuity degrees. Shown are three different sets of structurally analogous promiscuous or non-promiscuous kinase inhibitors. Inhibitors in the top and middle panel originated from ChEMBL. For these inhibitors, high-confidence activity data were available and their PD values are reported for confidence level 1 and 7. Cells containing different PD values are color-coded. Inhibitors in the bottom panel originated from PubChem and the number of kinases they were active or inactive against is reported (color-coded according to Fig. 3).

that ChEMBL also incorporates data from the PubChem BioAssay collection).

Promiscuity degrees. The PD values of all kinase inhibitors from PubChem ranged from 1 to 11, with mean of 1.1 (Fig. 1), which was comparable to the mean PD of 1.4 determined for ChEMBL inhibitors at the highest confidence levels. Hence, mean promiscuity was also very low for PubChem inhibitors. Remarkably, only 31 inhibitors were found to be active against at least five and only two against at least 10 kinases.

Activity versus test frequency. Fig. 3 reports the distributions of PubChem inhibitors over increasing numbers of human kinases against which they were tested and thus monitors the presence or absence of promiscuity at higher resolution. The majority of inhibitors was assayed against 10 to 20 different kinases. Over this range, mean PD values remained essentially constant at 1, while the number of kinases against which inhibitors were inactive steadily increased. Thus, most inhibitors from PubChem tested against multiple kinases were only active against a single kinase.

Exemplary inhibitors

Fig. 4 shows different sets of structurally analogous kinase inhibitors from ChEMBL (top, middle) and PubChem (bottom) having varying promiscuity degrees. In the first set (top), the promiscuous compound on the left had a PD value of 5 at confidence level 1 and a PD of 7 at level 7. Therefore, this compound was promiscuous at highest data confidence and displayed a moderate increase in promiscuity at the lowest confidence level. By contrast, its two structural analogs were only active against a single kinase at the highest and lowest confidence level. Furthermore, all three inhibitors in the second set (middle) were active against a single kinase at confidence level 1. Proceeding to confidence level 7, the PD value of only the inhibitor on the left increased from 1 to 3, whereas it remained constant for the other two analogs. Hence, these compounds were examples for selective inhibitors under varying data confidence. Moreover, the inhibitors of the third series from PubChem (bottom) were tested (from the left to the right) against 13, 10, and 18 different kinases. The promiscuous inhibitor on the left was found to be active against three of 13 kinases, while the other two analogs were only active against a single kinase. Thus, there were also selective within their experimental boundaries. These examples illustrate the prevalence of single-kinase inhibitors and also show that closely related analogs might have varying promiscuity degrees. In addition to data sparseness, assay variance and different assay formats might also influence promiscuity degrees, but it is evident that intrinsic promiscuity is far from being the rule among the kinase inhibitors investigated here.

Conclusions

In this study, we have – for the first time – analyzed the promiscuity of human kinase inhibitors at varying data confidence levels and taking test frequency information into account. The vast majority of publicly available kinase

inhibitors are directed against the ATP binding site. Since this site is largely conserved across the kinome, these types of inhibitors are often thought to be promiscuous, which has also become a paradigm for the use of kinase inhibitors in cancer treatment.

Over the past two years, the number of kinase inhibitors for which high-confidence activity data are available has more than doubled. Combining ChEMBL and PubChem as compound data sources, more than 141 000 kinase inhibitors with at least low-confidence activity data were obtained. For about 20% of these inhibitors, it was possible to determine target-based test frequency and the proportion of targets the compounds were active against. Thus, there was an excellent basis for re-visiting the issue of kinase inhibitor promiscuity *versus* selectivity by large-scale compound and activity data analysis, which has motivated our investigation.

Through systematic compound data mining only small subsets of highly promiscuous kinase inhibitors were identified. By contrast, the majority of inhibitors from medicinal chemistry and screening sources were only active against a single kinase. For more than 45 000 inhibitors with available high-confidence activity data, a mean PD of 1.4 was obtained. At decreasing data confidence levels, mean PD values of more than 92 000 kinase inhibitors remained low at around 2. Even in the absence of data confidence criteria, taking all activity information without restrictions into account, mean PD values were smaller than 4, but these values were biased by small numbers of highly promiscuous inhibitors (as shown by comparison of mean and median PD values). These findings were consistent with previous analyses focusing exclusively on high-confidence activity data. Accordingly, including increasing amounts of low-confidence activity data in promiscuity analysis did not lead to substantial increases in promiscuity degrees. Similarly, many kinase inhibitors that were tested in screening assays against 10 to 20 different kinase were only active against a single kinase and the mean PD value of all kinase inhibitors from screening sources was also close to 1.

Taken together, the results of our large-scale analysis show that promiscuity of ATP site-directed inhibitors of human kinases cannot be generalized. Rather, a differentiated view is required and potential selectivity of kinases inhibitors needs to be taken into consideration. Clearly, on the basis of currently available screening and activity data, the majority of kinase inhibitors are only active against one or at most two kinases at varying data confidence levels. These findings also have important implications for kinase inhibitor development. In many instances, it should be possible to chemically advance ATP site directed inhibitors and render them selective.

Conflicts of interest

There are no conflicts to declare.

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