Original Article Acquired rare recurrent EGFR mutations as mechanisms of resistance to Osimertinib in lung cancer and *in silico* structural modelling

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Abstract: A growing number of progression on Osimertinib among *EGFR*-mutated lung cancers represents a great challenge clinically. Our study aims to gain insights into novel mechanisms of acquired resistance to Osimertinib. We performed genomic studies on 2 large independent cohorts of lung cancer patients with progressed diseases on different tyrosine kinase inhibitors (TKIs). *In silico* modeling was used to study the structural mechanism of selected EGFR mutations. Compared with the 1st-TKIs-resistant group, *EGFR* mutations C797S/G, L718Q/V, L792F/H were significantly more enriched in the Osimertinib-resistant cohort, whose sensitivities to Osimertinib were successfully predicted. Importantly, a total of 14 low-frequency *EGFR* mutations were exclusively or significantly observed in the Osimertinib (G796S, V802F, T725M, Q791L/H, P794S/R). Analysis of pre-Osimertinib treatment samples of two patients supported that *EGFR* V802F and G796S were acquired during the treatment. In addition, *EGFR* G796S was predicted to be susceptible to gefitinib. This study represented the largest real-world data so far investigating Osimertinib resistance in *EGFR*-mutated lung cancer. We identified a collection of coexistent EGFR rare mutations and provided possible guidance for those patients who progressed on the first-line treatment of Osimertinib.

Keywords: EGFR recurrent mutation, Osimertinib, acquired resistance, *in silico* modeling, next generation sequencing

Introduction

Non-small cell lung cancer (NSCLC) represents one of the leading causes of cancer-related mortality worldwide [1], among which 10-50% patients bear activating mutations in epidermal growth factor receptor (*EGFR*), *e.g.* inframe deletions in exon 19 (Ex19del) or missense mutation in exon 21 (L858R) [2]. The first (1^{st}) and second (2^{nd}) generations of EGFR tyrosine kinase inhibitors (TKIs) suffer from considerable drug resistance, *e.g. EGFR* secondary mutation T790M [3], targeting which the third (3^{rd}) generation EGFR TKI Osimertinib (AZD9291) was approved by FDA in 2015 [4]. However, the majority of patients progressed on Osimertinib was found to develop newly acquired resistanct mutations, *e.g. EGFR* C797S, which destroys the covalent bonding with Osimertinib [5]. Recently, more and more *EGFR* mutations are discovered but their roles in Osimertinib resistance remain unknown [6].

Our early study of Osimertinib-resistant NSCLC patients revealed *EGFR* secondary mutations of L718 and L792, which are located in the p-loop and hinge of the ATP-binding site, respectively [7]. Further cellular experiments con-

firmed that both mutations were able to result in drug resistance against Osimertinib [8]. However, given the continual appearance of novel EGFR mutations and their extremely low frequency, using in vitro and/or in vivo experiments to predict drug sensitivity seems realistically forbidden. Recently, in silico methods, e.g. alchemical free energy calculation, have become promising in precision medicine for the patients carrying rare mutations [9]. For instance, Park et al. successfully applied a thermodynamic integration (TI) method to establish the role of the gatekeeper mutation on EGFR inhibitor selectivity [10]. Ikemura et al. reported that their established molecular dynamics (MD) simulation-based computational model, one of free energy perturbation (FEP) methods, predicted the diverse sensitivities of EGFR exon 20 insertion mutations to Osimertinib with surprisingly high consistency [11].

To explore the new resistant mechanisms, we analyzed the *EGFR* mutation profiles of 1,058 relapsed Chinese lung cancer patients after Osimertinib treatment, using targeted next-generation sequencing (NGS) with a comprehensive pan-cancer gene panel. Our study represents the largest *EGFR*-mutated lung cancer cohort so far investigating the Osimertinib resistance in a real-world setting. Furthermore, our *in silico* structural model was proved to be powerful and robust to predict the Osimertinib sensitivity of *EGFR* mutants.

Material and methods

Subjects

We retrospectively reviewed the genomic profiles of a cohort of 1058 lung cancer patients with acquired resistance to Osimertinib, a 3rdgeneration EGFR tyrosine kinase inhibitors (TKIs) (Osimertinib-resistant group), and another cohort of 1803 lung cancer patients, who have only received 1st-generation EGFR-TKIs prior to disease relapse (1st-TKI-resistant group). Written informed consent was collected from each patient upon sample collection according to the protocols approved by the ethical committee of each hospital.

DNA extraction, hybridization capture, and sequencing

Blood samples were collected from the above patients after the disease progression upon

drug treatment with Osimertinib or 1st generation TKIs. cfDNA was extracted using the NucleoSpin Plasma XS kit (Macherey Nagel). Purified DNA was qualified by Nanodrop2000 (Thermo Fisher Scientific) and quantified by Qubit 2.0 using the dsDNA HS Assay Kit (Life Technologies). Sequencing libraries were prepared using the KAPA Hyper Prep kit (KAPA Biosystems). A customized NGS panel sequencing was performed according to the previous report [12]. In brief, indexed DNA libraries were subjected to probe-based hybridization using IDT xGen Lockdown reagents (IDT, Coralville, IA) and Dynabeads M-270 (Thermo Fisher). Captured libraries were onbeads amplified with Illumina p5 and p7 primers in KAPA HiFiHotStartReadyMix (KAPA Biosystems). The final library was quantified by KAPA Library Quantification kit (KAPA Biosystems). Bioanalyzer 2100 (Agilent, Stanta Clara, CA) was used to determine the fragment size distribution of the final library. The targetenriched library was then sequenced on Illumina HiSeq4000 NGS platforms (Illumina) according to the manufacturer's instructions.

Sequence data processing and mutation calling

After demultiplexing, the sequencing data were subjected to Trimmomatic [13] for FASTQ file quality control (QC). Leading/trailing low quality (quality reading below 30) or N bases were removed. Qualified reads were mapped to the reference human genome hg19 using Burrows-Wheller Aligner (BWA-mem) [14]. Genome Analysis Toolkit (GATK) [15] was employed to apply the local realignment around indels and base quality score recalibration. PCR duplicates were removed by Picard. VarScan2 [16] was employed for the detection of single-nucleotide variations (SNVs) and insertion/deletion mutations with the recommended parameters [17]. We required minimum variant allele frequency of 0.3%, minimum variant supporting reads mapped to both strands of 3 for cfDNA samples. The resulted mutation lists were filtered through an internally collected list of recurrent sequencing errors on the same sequencing platform, which is summarized from the sequencing results of 400 normal blood samples. Specifically, if a variant was detected (i.e. ≥3 mutant reads) in >10% of the normal samples, it was considered a likely artifact and was removed. As a result, 480 (Osimertinib-resistant group) and 929 patients (1st-TKI-resistant group) bearing *EGFR* mutants were included in the following analysis, individually.

Computational modeling and simulation

We used co-crystal structures of wild-type EGFR-KD in complex with Osimertinib (PDB ID: 4ZAU) and Gefitinib (PDB ID: 4WKQ) individually, to eliminate the potential changes in the protein structure upon drug binding (Osimertinib or Gefitinib) or other non-targeted mutation(s). The strategy by Ikemura et al. [9] was applied to build the in silico model of mutant EGFR-drug complex that each mutation was introduced into wild-type EGFR structure after modeling disordered loops and flexible side chains, which was also used in other studies of mutant kinase-drug interaction [18-21]. MD simulation and end-point binding free energy calculations were implemented following the protocols previously reported [22]. The computational details were delineated in the Supporting Information. (Computational Methods Section) To confirm the reliability of in silico modeling, we also collected experimentally determined IC₅₀ (half inhibitory concentration) values for the specific mutants from our previous study [2], and compared the experimental results with our predictions.

EGFR mutation dynamics visualization

For *EGFR* mutation dynamics, we used Pyclone [23] and SCHISM [24] to estimate cancer cell fractions and clonal architecture, which was visualized using the fishplot R package (https://bmcgenomics.biomedcentral.com/articles/10.1186/s12864-016-3195-z).

Statistical analysis

The Fisher's exact test was applied to *EGFR* alterations with >1% prevalence in the Osimertinib resistant group. P<0.05 was considered to indicate a statistically significant difference. All statistical analyses were performed with R packages.

Results

Overview of patients' characteristics

Here, we conducted a multicenter record review of two large cohorts of lung cancer patients

who underwent genomic profiling at hospitals across China. The first cohort was comprised of a total of 1058 patients whose post-Osimertinib samples were included for analysis, regardless of prior TKI therapy other than Osimertinib. This cohort was delegated as the Osimertinib-resistant group and EGFR somatic mutations were detected in 480 patients. A total of 1803 patients who only received 1st-generation EGFR TKIs (e.g. Gefitinib, Erlotinib, Icotinib) prior to disease progression were allocated as the control group (1st-TKI-resistant group). Somatic EGFR mutations were detected in 929 patients of the 1st-TKI resistant group. A summary of the clinical characteristics of all patients was provided in Supplementary Materials (Table S1).

Identify EGFR mutations enriched in Osimertinib-resistant group

To identify EGFR mutations that may cause the resistance to Osimertinib rather than 1stgeneration EGFR TKIs, we compared the mutational profiles of Osimertinib resistant patients with 1st-TKI-resistant group. For this, alterations were filtered by two criteria: 1) significantly enriched (P<0.05) in the Osimertinib-resistant group compared to the 1st-TKIresistant group, and 2) rare but recurrent (N \geq 2) in Osimertinib-resistant group but not present in 1st-TKI-resistant group. We only focused on single amino acid substitution rather than complex mutations including large or small fragments insertion or deletion, for further MD simulation and structural analysis. The latter was believed to disturb the native protein structure more dramatically, therefore, it requires more extensive sampling and more accurate scoring for structural analysis.

Specifically, 7 mutations were found significantly enriched in the Osimertinib-resistant group compared with 1st-TKI-resistant group (**Figure 1**), including *EGFR* C797S mutation (20.6% vs. 0.4%, *P* value = 3.650818e-44), L718Q (4.6% vs. 0.2%, *P* value = 4.807101e-09), L792F (2.5% vs. 0.1%, *P* value = 2.229274e-06), L792H (1.9% vs. 0.2%, *P* value = 0.001574455), L718V (1.7% vs. 0.1%, *P* value = 0.0005172586), and G796S (1.3% vs. 0.1%, *P* value = 0.001531042), as shown in **Figure 1**. We also identified 13 recurrent mutations which were only observed in



Figure 1. *EGFR* mutations enriched for Osimertinib resistance. *EGFR* mutations significantly enriched in the Osimertinib-resistant cohort (P<0.01) and 13 rare but recurrent (n≥2) mutations observed only in Osimertinib-resistant cohort.

Osimertinib-resistant group but not in 1st-TKI-resistant group (**Figure 1**), *e.g.* G796S (1.2% in prevalence, 6 patients), V802F (0.6% in prevalence, 3 patients), Q791L/H (0.4% in prevalence, 2 patients), P794S/R (0.4% in prevalence, 2 patients) (<u>Table S2</u>).

Energetic analysis of EGFR mutations on Osimertinib binding

To investigate the effects on Osimertinib binding, we applied a molecular dynamics simulation-based model to predict the sensitivity of rare EGFR mutations to 1st- and 3rd-generation EGFR-TKIs. To achieve a good balance between computational efficiency and accuracy, we focused on the mutations around the ATP-binding site in EGFR kinase domain (KD). For this, 7 mutational sites around the ATPbinding pocket, that is within 10 angstrom from Osimertinib in the co-crystal structure (PDB ID: 4ZAU), were selected for further analysis, including EGFR L718Q/V, L792F/H/ P, T725M, Q791L/H, P794S/R, G796S, and V802F, among which the mutations at Q791 and P794, to the best of our knowledge, have not been reported previously.

The mutational sites that are located at the hinge region of EGFR KD, including *EGFR* Q791 (gatekeeper+1), L792 (gatekeeper+2), P794 (gatekeeper+4), G796 (gatekeeper+6), as shown in **Figure 2A**, were known to be essential for ATP binding [25]. In addition, both T725 and L718 are on the p-loop, which directly interact with small-molecule drugs (e.g. Osimertinib).

Moreover, V802 is located at the first helix right next to the hinge region. Notably, 5 mutations (e.g. G796S, V802F, T725M, Q791L/H, P794S/R) were found co-existing with C797S (Figure S1). However, the retrospective inspection of sequence alignments showed that 10 patients, involving Q791L/H, P794S/R, G796S, V802F individually, all harbored the concurrent mutation in trans with EGFR C797S (Figure S2), suggesting their potential roles in drug resistance against Osimertinib.

Our computational results showed that all of these EGFR mutations could reduce the binding energy of Osimertinib compared to that of wild-type (WT) EGFR evaluated by $\Delta\Delta G$ (Figure **2B** and **Table 1**). The most dramatic changes were observed with Q791H, L718Q, T725M (range from 4.84 to 6.74 kcal/mol, Table 1), then followed by P794R/S, G796S, and Q791L (range from 2.76 to 4.21 kcal/mol, Table 1), while V802F and L792F were found with minimal effects on Osimertinib binding (range from 1.09 to 1.52 kcal/mol, Table 1). Similar results were observed in the 1st-generation TKI, Gefitinib, except that G796S was predicted to increase the drug binding affinity ($\Delta\Delta G$: -2.37 kcal/mol, Figure 2B and Table 1).

Structural alterations underlying reduced drug binding affinity by EGFR mutations

To investigate the structural alterations underlying the reduced binding affinities by the above mutations, we compared the MD-simulated structural models with the native crystal structure and mainly focused on the drug binding as well as the protein context in the binding site. These EGFR mutations were predicted to interrupt the drug binding through various mechanisms including steric clash, hydrogen bond (HB) network alteration, and protein backbone strain elimination despite their close locations at the ATP-binding site.

Firstly, L718Q, L792F, and G796S all introduced steric clash which completely or partially impaired the key HB interaction with Osi-



Figure 2. A. Structural view of identified Osimertinib-associated EGFR resistant mutations around ATP-site, where the drug bind. The kinase domain of EGFR protein is shown in grey ribbons with selected mutations highlighted with filled red circles. The small-molecule drug Osimertinib is represented with magenta ball-and-stick model. B. Difference in the calculated binding energies between mutant and wild-type EGFR proteins for each drug (yellow: Osimertinib, grey: gefitinib), a positive value implies unfavorable binding while negative value suggests favorable binding.

 Table 1. Calculated energetic difference and root-mean-square deviation (RMSD) for small-molecule drugs in mutant and WT EGFR-KDs

| Mutation | L718Q | L792F | G796S | Q791H/L | P794R/S | V802F | T725M |
|--------------------------|-------|-------|-------|-------------|-------------|-------|-------|
| Osimertinib | | | | | | | |
| ΔΔG (kcal/mol) (vs. WT) | 4.86 | 1.09 | 3.76 | 6.74/2.76 | 4.21/2.90 | 1.52 | 4.84 |
| Ligand RMSD (Å) (vs. WT) | 2.163 | 1.724 | 2.670 | 2.297/1.743 | 1.331/1.835 | 2.086 | 2.271 |
| Gefitinib | | | | | | | |
| ΔΔG (kcal/mol) (vs. WT) | 6.22 | 1.46 | -2.37 | 2.35/-0.65 | 2.60/4.93 | 3.30 | 1.46 |
| Ligand RMSD (Å) (vs. WT) | 4.590 | 5.075 | 1.241 | 3.757/2.461 | 5.212/3.088 | 4.159 | 2.130 |

mertinib. Specifically, the bulkier and polar mutation (GIn) of L718Q is unfavorable to the binding of the aniline functional group in Osimertinib which results in steric hindrance thus destroying hydrophobic interaction with aromatic ring in comparison to the WT nonpolar residue (Leu), giving rise to a significant deviation of the drug (ligand RMSD: 2.163 Å) (**Figure 3A**; **Table 1**). This is consistent with the loss of *in vitro* activity as previously described [26]. Similarly, L792F produced a slightly bulkier side-chain substitution (Phe) at the hinge region, which leads to the clashes with the selectivity-determined methoxyl functional group (ligand RMSD: 1.724 Å) although the key hydrogen bond (HB) is largely maintained (**Figure 3B**; **Table 1**). This also explained the residue sensitivity of L792F to Osimertinib in the previous cellular experiments [27]. G796 is also located at the hinge region (gatekeeper+6), whose backbone forms direct contacts with the aniline core. The substitution (Ser) introduces a hydroxyl in the side-chain, which is against the hydrophobic phenyl group and produces steric hindrance that pushes drug significantly away from the binding site (ligand RMSD: 2.670 Å) (**Figure 3C**; **Table 1**). Interestingly, gefitinib was shown to



Figure 3. Structural insights into rare EGFR mutant variants bound with Osimertinib. Protein is shown in ribbon and stick for highlighted residues while drug molecule is represented with ball-and-stick model (white: crystal structure of WT EGFR, magenta: simulation structure of mutant EGFR). The selected mutations and induced structural changes are highlighted with red circles.

bind with EGFR G796S mutant even more favorably ($\Delta\Delta$ G: -2.37 kcal/mol, ligand RMSD: 1.241 Å) by forming additional HB with the hydroxyl group in mutated residue (<u>Figure S3</u>; **Table 1**).

Secondly, for the rare substitutions of Q791H/ L, a conserved HB network was observed between the side-chains of GIn791 and Lys852, Asp1012, Asp1014 in the WT EGFR, which however, was completely destroyed by these substitutions. This disturbed the binding of Osimertinib in the ATP-site, as suggested by the ligand RMSD values (Q791H: 2.297 Å, Q791L: 1.743 Å) (Figure 3D and 3E).

The third predicted mechanism of reducing drug binding affinity was the elimination of



Figure 4. Fish plots of *EGFR* somatic variants detected in Osimertinib-resistant patients. Arrows indicate different time points for sampling, e.g. after acquiring drug resistance against the 1st-generation EGFR-TKI (empty) and 3rd-generation EGFR-TKI (filled), as well as the pre-treatment (empty and dash line). Y-axis represents VAF (%) and X-axis shows the time progression.

native backbone strain by P794R/S which was introduced by the conserved hinge region [28]. The proline-produced backbone strain is believed to be important for the maintenance of the protein secondary and tertiary structure [29], therefore the mutations could be expected to destabilize the drug binding (ligand RMSD: 1.331 Å for P794R, 1.835 Å for P794S) (**Figure 3F** and **3G**).

Other structural changes were observed in the rare mutations V802 and T725 though they were located relatively far from the drug binding site. The simulation results showed that V802F mutation displaced the first helix right next to the hinge region compared to that in WT EGFR, as suggested by the atomic distance of backbone Ca (residue RMSD: 2.590 Å). This movement disturbed drug binding (ligand RMSD: 2.086 Å) (**Figure 3H**). Similarly, the bulkier and polar-to-hydrophobic mutation of T725M displaced the p-loop by moving towards the ATP-site, which resulted in the conflict with the drug (ligand RMSD: 2.271 Å) (**Figure 3I**).

Despite all rare mutations are located around the ATP-binding site, the large diversity of structural alterations underlying Osimertinib resistance present a great challenge to structure-based drug design of the next-generation EGFR-TKIs targeting ATP-binding site, which is known with a few common features, e.g. interaction with hinge region (with 1 or 2 hydrogen bonds), direct contacts with p-loop, and others. Encouragingly, as suggested by the increased binding affinity of Geftinib and G796S, a proper but unexpected drug repurposing can be achieved. This is of practical use in a clinical setting, which however, demands elaborate identification of cross-activity of different TKIs across a wide variety of resistance-associated mutant EGFRs.

V802F and G796S are acquired EGFR rare mutations coexistent with C797S

Considering the contribution of EGFR rare mutations to Osimertinib resistance, we were interested in exploring the mutation dynamics related to the drug resistance clinical course. With the retrospective inspection of our cohort, especially those with available electronic health record data, 2 patients who received Osimertinib treatment after acquiring 1st-TKI resistance, were characterized by the multiple samples collected at different time points, including pre- and post-treatment of Osimertinib. This allowed us to track EGFR genetic alterations within 2 years with four analyzed samples in the first patient (P1) and for over 2 years in the second patient (P2) with 6 samples.

As shown in **Figure 4A**, the coexistence of *EGFR* rare mutation G796S with C797S was observed in Patient 1 (P1). Specifically, P1 was initially treated with the 1^{st} -generation EGFR-TKI (lcotinib) for 1.5 years and then received the 3^{rd} -generation EGFR-TKI (Osimertinib) when the progressed disease (PD) occurred. The classic EGFR oncogenic driver mutation L858R and 1st-generation-TKI-resistance mutation T790M were detected in the

first sample (P1.1) collected upon Icotinib resistance. However, the sequential NGS analysis revealed multiple sub-clonal mutations (C797S, G796S, and L718Q) acquired after Osimertinib treatment. In addition, 3 different DNA-level variations were detected all leading to EGFR C797S mutation, including c.2389T> A (VAF: 22.4%), c.2390G>C (VAF: 22.8%), and c.2390GC>CA (VAF: 0.1%). In October 2018 (P1.3), mutational data indicated an increase of c.2389T>A (VAF: 37.6%) and a decrease of c.2390G>C (VAF: 8.3%) for EGFR C797S. The same mutations were also detected in P1.4 except for a considerable increase in the EGFR G796S (VAF: 5.7%), suggesting the contribution of G796S to Osimertinib resistance as well as the coexistence of dominant (C79-7S) and rare (L718Q, G796S) mutations. Of note, the 1st-generation EGFR-TKI, Gefitinib, was predicted to maintain the binding activity towards G796S mutant EGFR, suggesting the susceptibility to Gefitinib among patients with acquired resistance to first-line Osimertinib.

As shown in Figure 4B, the sequential occurrence of C797S after EGFR rare mutation V802F was observed in patient 2 (P2). A total of 6 samples from P2 were obtained for NGS analysis including treatment-naïve, post-lcotinib, and post-Osimertinib. In sample P2.2 (December 2016, pre-Osimertinib sample), EGFR T790M was detected with a VAF of 29.2%. With disease progression, another sample was analyzed in August 2017 (P2.3) and we were able to detect two new subclonal EGFR mutations, V802F and C797S with VAFs of 77.5% and 0.1%, respectively. The higher VAF of V802F suggested its potential role in drug resistance. In January 2018 (P2.4), mutational data indicated a dramatic increase in the EGFR C797S (VAF: 41.0%), accompanied by the significant decrease of EGFR V802F. Different from P1, a single DNAlevel alteration, c.2389T>A, was observed to cause EGFR C797S in P2. EGFR C797S lasted for the following time points (P2.5 and P2.6). In a word, our findings provided an additional view of mutation dynamics during the time course of EGFR-TKIs treatment.

Discussion

In this study, we provided both genomic and structural insights into the novel mechanism of

Osimertinib-resistant EGFR mutations using a very large cohort of lung cancer patients. Several rare mutations associated with drug resistance were identified, including T725M, 0791L/H, and P794S/R. All of them were characterized with a low frequency of <0.5%. Considering the recent approval of Osimertinib as the first-line treatment of metastatic NSCLC [29], the expanded use would inevitably result in more unrecognized drug-resistant mutations. This can be evidenced by the recent retrospective analysis of the FLAURA trial, which revealed that EGFR C797S was only present in 7% of patients acquiring resistance in the studied cohort [30]. Moreover, the landscape of drug resistance for Osimertinib is largely unknown. Our work has disclosed a collection of EGFR rare mutations enriched in Osimertinib resistant patients, which could serve as guidance in the clinical practice. Our findings also suggested EGFR C797S, as a dominant mutation, is likely to coexist with other rare mutation(s) (e.g. EGFR G796S and V802F) among a small proportion of patients progressed on Osimertinib treatment, which together, contribute to the drug resistance. The coexistence of EGFR rare mutation(s) with C797S observed in the present study suggested the mechanism of resistance to Osimertinib might be heterogeneous.

Considering the extremely low frequency of EGFR rare mutation and the continual appearance of novel mutations, experimental evaluation of drug sensitivity is realistically forbidden in a clinical setting. Encouragingly, in silico methods hold promise in providing alternative solutions. Several studies turned to the alchemical free energy calculation, e.g. FEP and TI, and proved the power in precision medicine for the patients carrying rare mutations [9-11]. However, the extremely high computational cost largely prevents practical use, which demands the assistance of supercomputers or clusters. In contrast, the endpoint method is shown to achieve a good balance between efficiency and accuracy [26]. In the present work, we predicted both EGFR L718Q and L792F mutants are able to reduce the binding affinities of Osimertinib and Gefitinib, and L7180 renders more resistance to both drugs than L792F. These results are in agreement with our previous experimental findings that both EGFR mutants (L718Q and L792F) render

increased $\mathrm{IC}_{_{50}}$ values for two drugs compared with EGFR without the specific mutation, as well as L718Q presents much higher IC₅₀ values than L792F for Osimertinib (L718Q: >1000 nM, L792F: 72.41 nM) and Gefitinib (L718Q: 500 nM, L792F: 31.25 nM), respectively [2]. Our results suggested the endpoint free energy calculation, e.g. MM/PBSA, is able to delineate drug sensitivity across a large diversity of EGFR mutations by simply estimating and comparing the binding free energies. Herein, we only focused on the single point mutation rather than more complex ones, e.g. small/large fragment insertion/deletion, which is believed to disturb the native protein structure more dramatically, and therefore, demands much more expensive computations.

The atomic-level MD simulations could provide structural insights into the mechanism of drug resistance. Despite the very close locations within or around ATP-binding site, many different roles were still seen for the studied rare mutations, e.g. introducing steric clash and impairing key HB interaction with hinge region (L718Q, L792F, G796S), disturbing protein native HB network (Q791H/L), eliminating imposed backbone strain in the native structure (P794R/S), and other effects. These together, present a great challenge for structure-based drug design, because most of small-molecule TKIs are known to target ATPbinding site, the large diversity of structural mechanisms observed for rare mutations in the present study suggest an unrealistic optimism in developing EGFR-TKI with ability to overcome all these resistant mutation effects simultaneously. Notably, our computational study found that compared with EGFR WT, Gefitinib even prefers to bind with EGFR G796S which shows resistance against Osimertinib, indicating the potential therapeutic strategy for this drug-resistant mutation. However, a proper but unexpected drug repurposing can only be achieved by elaborating the comprehensive profiles of cross-activity of different TKIs across a wide variety of resistant mutations, where in silico methods would be of great practical use.

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Disclosure of conflict of interest

Ran Cao, Qiuxiang Ou, Yutong Ma, Hua Bao, and Xue Wu are employees of Geneseeq Technology Inc., Toronto, Canada. Yang Shao is an employee of Nanjing Geneseeq Technology Inc., Nanjing, Jiangsu, China.

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Computational methods

In-silico modeling of EGFR-TKI complex

The crystal structure of human EGFR kinase domain (KD) in complex with osimertinib, the third-generation tyrosine kinase inhibitor (3^{rd} -TKI), was obtained from the Protein Data Bank (PDB ID: 4ZAU). The disordered loops and flexible side chains were modeled using Modeller v9.2 program (1). As a comparison, we also used the crystal structure of human EGFR-KD in complex with gefitinib (1^{st} -TKI, PDB ID: 4WKQ) as our computational control. For the selected mutations, each was introduced into the crystal structure of wild-type (WT) EGFR with fully structural optimizations to eliminate any potential induced steric clashes.

System setup and MD simulation

MD simulations were performed with AMBER12.0 suite (2). AMBER99SB force field was applied for the protein, and the general Amber force field (GAFF) was applied for the small molecules. ANTECHAMBER was used for calculating the AM1-BCC charges of small molecules. All system setups were performed in TLEAP and simulations were carried out using SANDER, during which 3 stages of minimization were performed in the gas phase, and then followed by the addition of a 30 Å water cap based on the geometric center of binding site. After 200 ps of equilibration of solvents, a production run of 5 ns was performed on the whole system at 300 K with a time step of 2.0 fs. All residues including solvents beyond 12 Å of small molecules are fully frozen.

Endpoint binding energy calculation

Molecular Mechanics Poisson-Boltzmann Surface Area (MM-PB/SA) calculation was carried out following the protocol previously reported (3). Generally, 100 snapshots were evenly extracted from last 1 ns MD simulation. The interaction energy between protein and small molecules as well as the internal energy were calculated with the SANDER. As for the solvation energy, the polar contribution is calculated using PBSA with PARSE radii, while the nonpolar part is estimated proportional to the solvent-accessible surface area. The difference in the binding energies can be estimated with the formula: $\Delta\Delta G = \Delta G_{Mutant} - \Delta G_{WildType} = (G_{complex} - G_{receptor} - G_{ligand})_{WildType}$.

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| Characteristics | Group I (n = 480) Osimertinib_resist | Group II (n = 929) 1 st -TKI-only_resist | p-value |
|---------------------------|---|--|---------|
| Male (%) | 201 (41.9%) | 380 (40.9%) | 0.7322 |
| Female (%) | 279 (58.1%) | 549 (59.1%) | |
| Age at diagnosis (Median) | 24-87 (61) years | 21-96 (62) years | >0.01 |
| | | | |

Table S1. Clinical characteristics of patients with identified EGFR somatic mutations

Table S2. Clinical characteristics of patients with identified EGFR somatic mutations

| Mutations | • | l (n = 480) tinib_resist | Group 1 st -TKI | p-value | |
|-----------|------------|-----------------------------|-------------------------------|-----------------|----------|
| | Prevalence | No. of patients | Prevalence | No. of patients | |
| T790M | 0.385 | 185 | 0.406 | 377 | |
| C797S | 0.206 | 99 | 0.004 | 4 | 3.65E-44 |
| L718Q | 0.046 | 22 | 0.002 | 2 | 4.81E-09 |
| L792F | 0.025 | 12 | 0.001 | 1 | 2.23E-06 |
| R252H | 0.021 | 10 | 0.011 | 10 | |
| V148M | 0.019 | 9 | 0.015 | 14 | |
| L792H | 0.019 | 9 | 0.002 | 2 | 1.57E-03 |
| A750E | 0.019 | 9 | 0.016 | 15 | |
| L718V | 0.017 | 8 | 0.001 | 1 | 0.000175 |
| C797G | 0.015 | 7 | 0.001 | 1 | 0.000517 |
| G796S | 0.013 | 6 | 0.001 | 1 | 0.001531 |
| C240Y | 0.013 | 6 | 0.011 | 10 | |
| V904I | 0.010 | 5 | 0.003 | 3 | |
| R958H | 0.010 | 5 | 0.009 | 8 | |
| A822T | 0.010 | 5 | 0.004 | 4 | |
| A1102T | 0.010 | 5 | 0.005 | 5 | |
| V1097I | 0.008 | 4 | 0.001 | 1 | |
| R451C | 0.008 | 4 | 0.001 | 1 | |
| V802F | 0.006 | 3 | 0.000 | 0 | |
| E711K | 0.006 | 3 | 0.000 | 0 | |
| V323I | 0.004 | 2 | 0.000 | 0 | |
| T725M | 0.004 | 2 | 0.000 | 0 | |
| P772T | 0.004 | 2 | 0.000 | 0 | |
| L792P | 0.004 | 2 | 0.000 | 0 | |
| K860I | 0.004 | 2 | 0.000 | 0 | |
| G729V | 0.004 | 2 | 0.000 | 0 | |
| G1103D | 0.004 | 2 | 0.000 | 0 | |
| E204K | 0.004 | 2 | 0.000 | 0 | |
| A1155T | 0.004 | 2 | 0.000 | 0 | |
| Q791L/H | 0.004 | 2 | 0.000 | 0 | |
| P794S/R | 0.004 | 2 | 0.000 | 0 | |

Mechanism of Osimertinib resisitance



Figure S1. Highlighted EGFR and other driver genetic mutations detected among 12 selected lung cancer patients. The original oncogenic mutations (L858R and 18del) are colored in black, the major resistant EGFR mutations at T790 and C797 are colored in rosybrown (1st-TKI resistant) and red (3rd-TKI resistant), the minor resistant EGFR mutations at L718, L792, G796, Q791, P794, V802, T725 are colored in skyblue, while mutations in other selected genes are colored in yellow (TP53), orange (PIK3CA), cyan (RET), forest green (BRAF).





V802F



T725M



Q791L







P794R



Figure S2. Sequence alignment for the selected patients with a focus on the EGFR resistant mutations around ATPbinding site.



Figure S3. Structural insights into the effects of EGFR-G796S mutation on gefitinib binding by comparing the complex structure of WT and mutant EGFR bound with drug. Protein is shown in ribbon and stick for highlighted residues while drug molecule is represented with ball-and-stick model (white: crystal structure of WT EGFR, magenta: simulation structure of mutant EGFR). The hydrogen bonds were shown in cyan and orange lines.