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Allostery in its many disguises: from theory to applications

Supplementary Material

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Figure S1: Population shift between the inactive and active states of Ras-GDP/GTP.
(Ruth Nussinov, Chung-Jung Tsa and Hyunbum Jang)

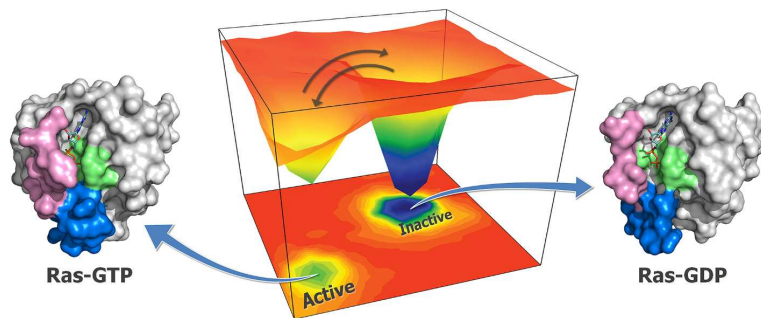


Illustration of the population shift phenomenon related to the authors' recent work on oncogenic mutants of the KRas4B isoform. Shown is the free energy landscape representing active Ras-GTP and inactive Ras-GDP states. Ras can be activated by guanine nucleotide exchange factors that exchange GDP with GTP. The active GTP-bound Ras can be inactivated through GTP hydrolysis by GTPase-activating protein (GAP).

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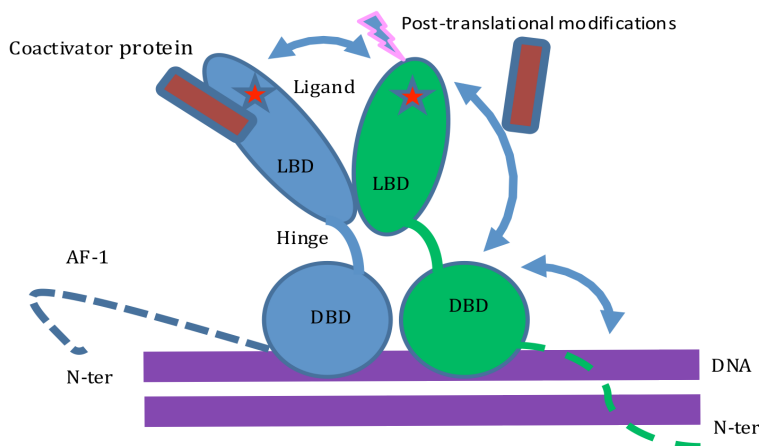


Figure S2: Allosteric regulation in nuclear receptors

(R.H. Stote, J. Eberhardt, Y. Chebaro, A. Dejaegere)

Schematic illustration of the domain structure of nuclear receptor proteins. This structure comprises the AF-1 N-terminal domain, DNA binding domain (DBD), hinge region, C-terminal ligand binding domain (LBD). Sources of allosteric signals include ligand binding and post-translational modifications, which can lead to allosteric signal

transmission to the DBD affecting DNA binding. Ligand binding by one receptor can affect ligand binding by the second receptor and the DNA sequence can modulate co-activator binding.

Computational tools have been instrumental in understanding allosteric regulation of NRs [1],[2]. An early study on the PPAR γ LBD identified networks of dynamically coupled amino acids that link the ligand binding pocket to the activation helix H12 and the heterodimer interface [3]. A more recently community network analysis has been applied to a larger PPAR architecture [4].

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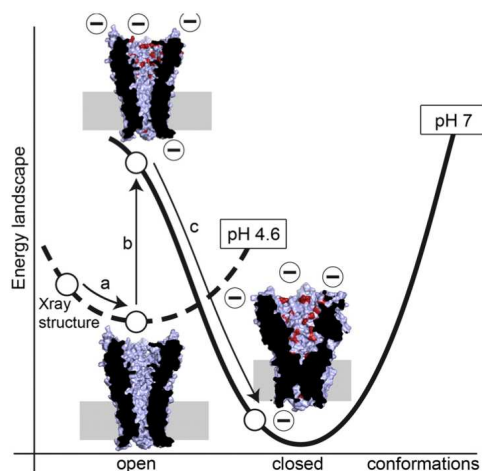


Figure S3: The Activation/relaxation MD approach to explore allosteric transitions of ligand-gated ion channels with atomic resolution: consequences for drug design
(Marco Cecchini and Jean-Pierre Changeux)

Schematic illustration of the activation/relaxation MD approach to allosteric transitions in ligand-gated ion channels. Starting from the X-ray structure of the bacterial receptor GLIC at pH 4.6 (open), a brief equilibration is carried out (a). Then, an instantaneous increase in pH is modeled (b), followed by the relaxation towards a closed conformation (c). The curves with broken and plain lines represent energy landscapes for pH 4.6 and 7.0, respectively. The protein's surface is represented in light blue with residues

changing charge during the pH jump in red. Adapted with permission from Ref. [2].

The all-atom based molecular dynamics activation/relaxation approach described in the main text, relies on the hypothesis that allosteric proteins spontaneously populate a number of discrete conformational states in reversible equilibrium and that a conformational selection or *shift* of conformer population takes place upon ligand binding/unbinding [1]. In this view, modeling the addition (removal) of the agonist to (from) its binding site is expected to trigger a “vertical excitation” of the protein, whose structural relaxation by unbiased MD would capture the spontaneous transition to a distinct physiological state..

Lastly, knowledge of the transition path between pairs of physiologically relevant states at atomic resolution enables detailed analysis of the dynamics of the ligand-binding pockets. These include sites for orthosteric ligands and allosteric modulators, which are most often located at subunit or domains interfaces and typically open and close during the functional transitions between resting, active (and desensitized) states. Such analyses can be exploited for the rational design of positive or negative allosteric modulators [1], paving the way to new pharmacological strategies.

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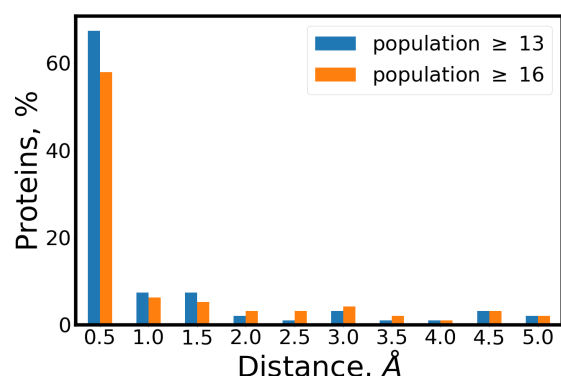


Figure S4: Structural Origins of Cryptic Binding Sites

(Dzmitry Padhorny and Dima Kozakov)

This figure shows the distribution of distances between the cryptic binding site and the closest FTsite probe cluster with above threshold population in the CryptoSite dataset [1]. Analysis is performed based on apo structures of the proteins. In the FTsite approach, clusters of size 13 and more are considered to be binding site predictors, which means that FTsite is able

to correctly identify the vast majority of known cryptic binding sites. In a similar manner, clusters with more than 16 probes pinpoint “druggable” binding sites which could be targeted with sufficient affinity, thus showing that more than 80% of known cryptic binding sites are druggable.

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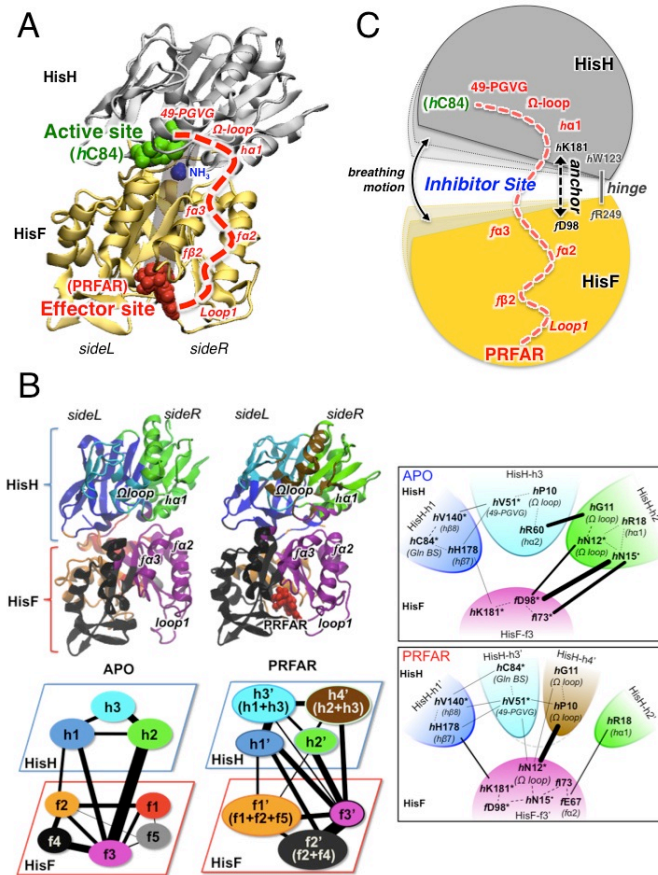


Figure S5: Community network analysis elucidating allostery regulation in the imidazole glycerol phosphate synthase (IGPS) protein
(Ivan Rivalta)

(A) Tertiary structures of *Thermotoga maritima* IGPS complex from *Thermotoga maritima*, with active site in the HisH protein (gray), allosteric site in the HisF protein (yellow) and ammonia (blue) channel (gray tube). IGPS catalyzes the hydrolysis of glutamine (Gln) at the *h* active site, upon binding of the effector PRFAR to HisF, >25 Å away. The allosteric pathways involve secondary structures (in red) on the IGPS right side (sideR). (B) Color-coded optimal community network of apo (left) and PRFAR-bound (right) complexes. The apo communities that contribute the most to the PRFAR communities are reported in brackets. Connectivity line-widths are proportional to the inter-communities total betweenness (ITB) with connections relevant to IGPS allostery depicted in the right panels. Fully and partially conserved residues are indicated with (*) and (•), respectively. Secondary structures are in brackets. (C) The allosteric communication involves structural changes induced by PRFAR

binding, affecting the HisF/HisH breathing motion that has been targeted by small molecule inhibitors that bind at the interface. Community network analysis [2,3] was applied to derive optimal community networks of the apo and effector PRFAR-bound IGPS complexes (B), indicating that effector binding alters significantly the information flow between communities (thickness of the links between communities in (B)). The findings from these studies have been corroborated by measurements of NMR chemical shifts and were shown to be consistent with the expected inactive-to-active allosteric transition in IGPS. This studies have stimulated both single site mutagenesis experiments[4] and allosteric inhibitors design [5] (see (C)).

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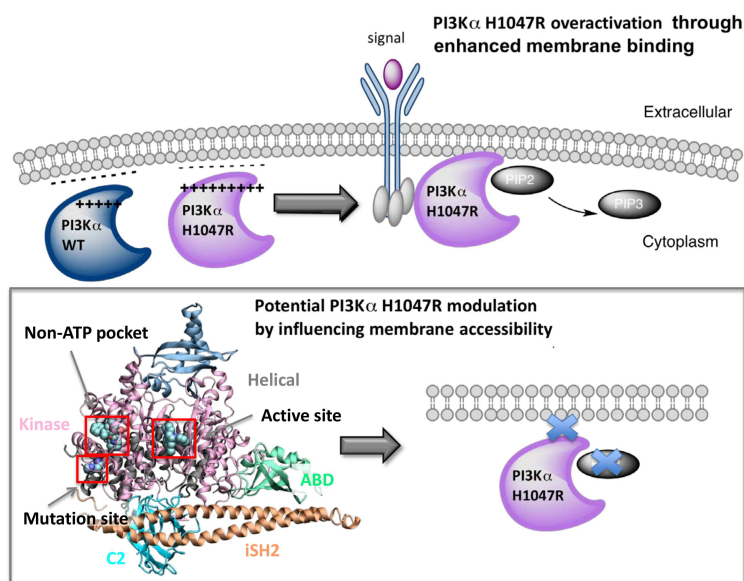


Figure S6: Allosteric modulation in drug discovery: The PI3K α paradigm
(Paraskevi Gkeka)

Schematic illustration of the proposed allosteric mechanism phosphoinositide 3-kinase alpha (PI3K α) [1,2].

PI3Ks are lipid kinases that play key roles in many fundamental biological processes, including cell growth, proliferation, differentiation, motility, survival, and metabolism. Among the different PI3K isoforms, PI3K α is the most important as it plays a pivotal role in cell proliferation in response to growth factor-tyrosine kinase pathway activation [3]. PI3K α is a

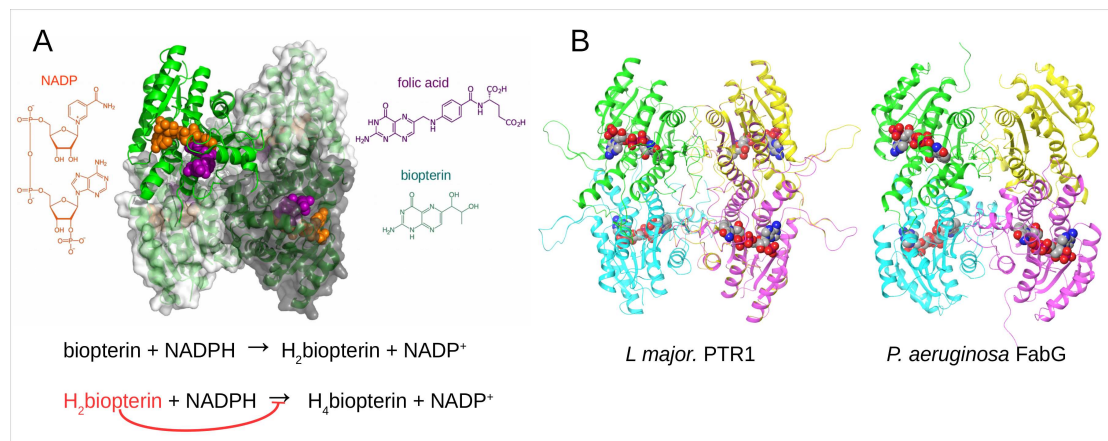
heterodimer that contains a p110 α catalytic subunit and p85 α regulatory subunit [1], and is a target of particular pharmacological interest for anti-cancer drug development [3,4]. Currently, two PI3K α inhibitors, Alpelisib and MLN1117, are in Phase III trial for patients with HR+/HER2-advanced breast cancer and Phase II trial in patients with advanced solid tumors, respectively (<https://clinicaltrials.gov>, November 2017). Despite the advances in targeting the PI3K α isoform, a major challenge remains because of the highly conserved ATP binding sites of lipid kinases. Based on molecular modeling techniques and surface plasmon resonance experiments, a multifactorial model of the over-activation mechanism of the most common *PIK3CA* mutant (H1047R), was proposed [1]. Using this model a new binding pocket distinct from the active site, and close to residue 1047, was identified. This pocket was further explored for potential allosteric modulation of the H1047R PI3K α mutant [2]. Positional covariance, protein fluctuation analyses showed that the main areas involved in protein conformational changes upon ligand binding to the non-ATP pocket are the membrane binding domains. PI3K α mutant activity could therefore, be modulated not in terms of the active site activity, but by altering the dynamics of protein-cell membrane interaction and subsequent substrate retrieval. Thus, the non-ATP pocket could potentially be used for the discovery of a selective inhibitor of protein-membrane interactions tailored for the H1047R mutant protein. Such an inhibitor could act by constraining PI3K α membrane-binding motifs, influencing membrane accessibility and subsequent substrate availability, representing a promising alternative or complementary strategy for allosteric PI3K α modulation.

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Figure S7: Allosteric modulation of pteridine reductase 1, a member of the short-chain dehydrogenases/reductases family

(Joanna Panecka-Hofman and Rebecca C. Wade)



Displayed are experimental and modeled structures of Pteridine reductase 1 (PTR1, EC 1.5.1.33), a folate pathway enzyme unique to trypanosomatid parasites. **(A)** PTR1 homotetramer structure (*T. brucei*, pdb: 3bmc), and the chemical structures of the main substrates and the NADP cofactor, and the reaction catalyzed by PTR1 with biopterin as an exemplary substrate (substrate inhibition shown in red). **(B)** Structures of PTR1 (a model based on pdb: 1e92) and FabG (pdb: 4ag3). The proteins are shown in cartoon and molecular surface representation and NADP as van der Waals spheres (only heavy atoms are shown). PTR1 catalyzes the reduction of folate and biopterin with NADPH cofactor [1]. It belongs to a large family of short-chain dehydrogenases/reductases (SDRs), which feature diverse sequences and very diverse functions [3], but have a highly conserved Rossmann fold [3]. PTR1 is a potential target for anti-parasitic drugs, but existing inhibitors require optimization [2]. Targeting allosteric sites of PTR1 is therefore an attractive alternative.

PTR1 is inhibited by semi-products of its catalytic reaction (substrate inhibition) **(A)** at substrate concentration of ~10-mM levels, and may involve an allosteric mechanism [1]. Cooperativity between the binding sites is also postulated for other SDR enzymes, including homotetrameric bacterial 3-ketoacyl-[acyl-carrier-protein] reductase (FabG, **B**) [4], and allosteric inhibitors binding at the inter-subunit interfaces were reported for FabG [5]. The FabG tertiary and quaternary structure is similar to that of PTR1, despite low sequence identity (~29% for *L. major* PTR1 and *P. aeruginosa* FabG **(B)**). The possibility of allosteric regulation in PTR1 was investigated using normal mode analysis (NMA), indicating significant concerted movements of the substrate loops flanking the active site. In addition, Rotamerically Induced Perturbation simulations (RIP [6]), revealed flexibility hot-spots in the PTR1 homotetramer structure, and suggested dynamical allosteric couplings between distant protein residues.

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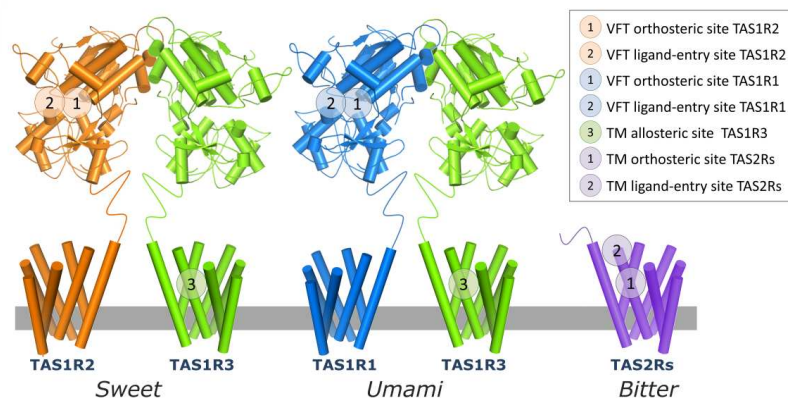


Figure S8: Allosteric modulation of taste GPCRs

(Antonella Di Pizio and Masha Y. Niv)

The Figure illustrates the ligand-binding sites of sweet (TAS1R2/TAS1R3, in orange and green), umami (TAS1R1/TAS1R3, in blue and green) and bitter (TAS2Rs, in violet) taste receptors, members of the G-protein coupled receptors (GPCRs) family.

The odorant or olfactory receptors (ORs) represent the largest family of GPCRs, with over 1000 members in mice and ~400 in humans. Sweet (elicited by sugars) and umami (elicited by glutamate) taste modalities are mediated via TAS1Rs, and bitter (elicited by caffeine, quinine, strychnine and hundreds of other compounds), is mediated by the TAS2Rs family that counts 25 subtypes in humans. Only the crystal structures of the extracellular VFT regions of the fish Tas1r2/Tas1r3 heterodimer are currently available [1], but iterative combinations of simulation and experiment have been successfully used for delineating binding modes of tastants, and for predicting additional ligands [2]. TAS1Rs (Class C GPCRs) contain an extracellular VFT as an N-terminal domain, linked to the TM domain via a short cysteine-rich (CR) domain. VFT, the orthosteric ligand-binding domain, consists of two lobes that can assume an open (inactive) or closed (active) conformation. Umami taste is mediated by a heterodimer composed of TAS1R1 and TAS1R3, and sweet is sensed by a heterodimer of TAS1R2 and TAS1R3.

The Sweet receptor, TAS1R2/TAS1R3 heterodimer is activated by diverse compounds, ranging from low molecular weight sweeteners to sweet proteins. The umami receptor, TAS1R1/TAS1R3 heterodimer, specifically responds to glutamate. Integrated computational approaches and mutagenesis studies identified the binding site for sugars in the VFT domain of TAS1R2, and of glutamate in the VFT domain of TAS1R1, as reviewed in [2]. An allosteric binding site for small molecules has also been identified in the TM domain of TAS1R3. Cyclamate and lactisole bind to an allosteric site in the TM domain, and act as allosteric enhancer and negative modulators, respectively, for both sweet and umami receptors [2]. In addition, multiple binding sites for different ligands have been identified on the sweet taste receptor. For example, small agonist molecules, such as sucralose, saccharin, aspartame and neotame, bind to the orthosteric binding site in the TAS1R2 VFT domain, but positive allosteric modulators, sweet taste enhancers, were found to bind at the entrance of the VFT domain of the TAS1R2 (ligand-entry site).

Allosteric regulation in bitter receptors is less well understood. TAS2Rs are classified as Class A GPCRs, although the sequence similarity is very low (13-29% for the TM domains). Despite low sequence identity the orthosteric binding pocket of bitter taste receptors coincides with the canonical binding site of Class A GPCRs. In addition, an extracellular or vestibular site has been described to be transiently involved in the binding process [3].

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