

Molecular and functional profiling of G α i as an intracellular pH sensor

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Heterotrimeric G proteins (G α , G β and G γ) act downstream of G-protein-coupled receptors (GPCRs) to mediate signaling pathways that regulate various physiological processes and human disease conditions. While human G α i and its yeast homolog Gpa1 were previously postulated to function as intracellular pH sensors, the pH-sensing capabilities of G α i and the underlying mechanism remain to be established. Our research shows that variations in pH significantly affect the structure and stability of G α i-GDP. Specifically, at the lower end of the physiological pH range, the protein undergoes an order-to-disorder transition due to the loss of electrostatic interactions within the G α i Switch regions, resulting in a reduction in agonist-mediated G α i-G $\beta\gamma$ release. Further, we identified key residues within the G α i Switch regions that form the pH-sensing network. Mutation of these residues in G α i gives rise to ‘low pH mimetics’ that abolish pH-dependent thermostability changes and reduce G α i-G $\beta\gamma$ release. Overall, our findings suggest that pH-sensitive structural changes in G α i impact the agonist-mediated dissociation of G $\beta\gamma$, which is essential for proper signaling.

Within the complex milieu of living cells, intracellular pH (pH_i) is maintained within a narrow range, as even small changes in pH can affect a myriad of cellular processes including membrane potential, ion transport, cellular growth and metabolism^{1–3}. Unsurprisingly, disruptions in pH regulation can contribute to the development of pathological conditions such as ischemic heart disease, cancer and neurological disorders^{4–6}. While various proton pumps and transporters play a role in regulating the flow of protons across the membrane to uphold pH_i homeostasis, there are intracellular proteins (termed pH sensors) that sense and transmit pH signals, thus orchestrating the regulation of biochemical processes. Among these pH sensors are signal-transducing proteins.

Notably, a study by Isom et al.³ provided evidence that a subset of signal-transducing heterotrimeric G proteins may serve as intracellular pH sensors³. These membrane-associated proteins form a

heterotrimeric complex (G α , G β and G γ) and act downstream of G-protein coupled receptors (GPCRs). GPCRs represent the most extensive group of membrane proteins that are targeted by approved drugs. They play a crucial role in orchestrating the majority of cellular responses to hormones and neurotransmitters through several signaling pathways mediated by different isoforms of G protein (G α i, G α s, G α 12/13 and G α q) that receive and transduce signals through diverse pathways^{7,8}. The G β and G γ subunits associate to form a G $\beta\gamma$ heterodimer³, whereas the G α subunit binds GDP or GTP and catalyzes GTP hydrolysis. The G α subunit is comprised of two distinct domains: a helical domain and a Ras-like domain. Within the Ras-like domain, there are three key ‘Switch’ regions, namely SW-I, SW-II and SW-III, which play an essential role in its activity. In the GDP-bound state, which is an inactive form of G α protein, these Switch regions exhibit dynamic behavior. In contrast, in the GTP-bound state (the active form

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of G α), they become more structured and less dynamic⁹. In the GDP-bound state, the G α subunit is associated with the G $\beta\gamma$ complex. Upon GPCR-stimulated G α GTP-loading, the G $\beta\gamma$ subunits dissociate from G α -GTP and along with Ga, promote activation of downstream signaling pathways.

Isom et al.³ developed a computer algorithm, pHinder, which predicted that both the mammalian G α i isoform and yeast homolog Gpa1 contain a core of residues between the Ras-like and helical domains that may promote pH-sensing properties³. In support of this prediction, both proteins showed pH-dependent changes in thermostability over a pH range from 5.5 to 8. They also found that Gpa1 undergoes phosphorylation under acidic conditions to attenuate pheromone-dependent stimulation of mitogen-activated protein kinases in the yeast³. While these findings, taken together suggest that mammalian G α i may function as a pH sensor, the molecular mechanism, and the biological consequence of pH sensing through G α i remain unknown.

To expand on past observations, we characterize the pH-dependent biochemical and structural/dynamic properties of G α i-GDP, elucidate the underlying pH-dependent electrostatic network, and assess the functional consequences on cellular G α i activity. We find that the structure, stability, and dynamics of G α i in the GDP-bound state are highly dependent on pH over the physiological pH range due to a pH-responsive network within the Ras-like domain. These findings differ from a previous report where larger pH-dependent changes in thermostability were observed for the GTP-bound state with ionizable residues predicted to lie at the interface between the Ras-like and helical domains³. Our NMR, biophysical and computational analyses indicate that changes in the ionization state of residues within the pH-sensing network promote a disorder-to-order transition in G α i over the physiological pH range, to populate a less ordered state that enhances G α i-G $\beta\gamma$ association in HEK293 cells at the lower end of the physiological pH range. Identification of the pH-sensing network allowed for the identification and generation of low pH mimetics that reduce pH-dependent G $\beta\gamma$ release from the G α i-G $\beta\gamma$ complex. Of note, the G α i Switch III region plays a key role in pH-sensing, suggesting a potential role for this understudied Switch region in agonist-mediated G $\beta\gamma$ release, a key step for both G α i and G $\beta\gamma$ activation. Taken together, our studies indicate that G α i undergoes a pH-dependent disorder-to-order transition that modulates G α i-G $\beta\gamma$ interactions and G α i activation over the physiological pH range.

Results

pH modulation of G α i-GDP structure and dynamic properties

Earlier investigations into mammalian G α i demonstrated changes in structure and stability in response to pH variations³. While these findings suggest a potential role for G α i as a pH sensor³, the molecular basis and functional relevance have yet to be established. Herein, we apply comprehensive and multidisciplinary approaches to evaluate how pH changes in the physiological range affect structure, stability and G α i activity in vitro and in cells. We first conducted CD experiments on G α i-GDP to monitor pH-dependent changes in thermal unfolding, stability and secondary structure. Further, to monitor thermal unfolding and stability, we collected CD spectra at 222 nm as a function of pH and temperature. As shown in Fig. 1A, a striking and gradual increase in thermal stability (ΔT_m ~ 25°C) of G α i-GDP was observed over the pH range from 5 to 7.3. Notably, the thermal unfolding transition for G α i-GDP appears cooperative at low pH but shifts to a multi-state unfolding transition over the physiological pH range (Fig. 1A). G α i contains two subdomains, a Ras-like and helical domain. To evaluate differential unfolding at higher pH (above pH 7.1), we conducted CD scans (200–250 nm) for G α i-GDP as a function of temperature at pH 7.2. As shown in Figs. 1B and 1C, CD spectra revealed a significant reduction in alpha-helical propensity, but not the beta-sheet propensity as the temperature is raised from 45–55°C. These

findings indicate that the helical domain melts first, followed by the Ras-like domain, with the Ras-like domain showing the greatest change in thermal stability at higher pH.

Given the enhanced stability observed at higher pH, we employed intrinsic tryptophan fluorescence experiments using tryptophan (W211) in the SW-II region as a fluorescence probe to monitor differences in solvent exposure as a function of pH, which indirectly indicates pH-dependent switch conformational changes. It has previously been shown that the more dynamic and less ordered GDP-bound state of G α i promotes enhanced exposure of W211, resulting in a fluorescence decrease relative to that of the G α i-GTP state^{9–11}. As shown in Fig. 1D, intrinsic G α i-GDP tryptophan fluorescence increases over the pH range from 5 to 7.2, suggesting that higher pH promotes decreased solvent accessibility, possibly due to enhanced interactions and structural order, consistent with the greatly enhanced stability observed by CD. Taken together, these results support a disorder-to-order transition at higher physiological pH for G α i in its GDP-bound form.

To further probe whether a pH-dependent disorder-to-order transition occurs in G α i-GDP, we conducted 2D NMR analyses. For these studies, we prepared ¹⁵N enriched G α i-GDP and collected a 2D ¹H-¹⁵N Heteronuclear Single Quantum Coherence (HSQC) NMR spectra over a physiologically relevant pH range (6.4–7.6). Enrichment with ¹⁵N allows the detection of backbone and sidechain NH resonances within G α i and provides a site-specific probe for every amino acid except proline. The 2D HSQC overlay of G α i-GDP at pH 6.4, 6.8 and 7.2 is shown in Fig. 2A, with an HSQC overlay comparing pH 6.4 versus pH 7.6 shown separately in Supplementary Fig. 1. Spectra acquired at lower pH show significant chemical shift changes, line broadening and loss of several peaks in comparison to those obtained at higher pH values, suggesting the protein is more dynamic at lower pH (Fig. 2A). This data correlates well with CD and tryptophan fluorescence data which suggests a less thermostable structure of G α i-GDP at lower pH. The pH-dependent HSQC changes observed are consistent with an earlier report which showed extensive broadening of NH peaks in the ¹H-¹⁵N HSQC spectrum of G α i-GDP at pH 6 relative to pH 7¹². Residue-specific chemical shift perturbation and peak intensity changes associated with changes in pH between 6.4 and 7.2 are plotted in Figs. 2B and 2C, respectively. Most pH-dependent spectral changes are localized to α 1, α 5, β 1, β 2 and the β 2- β 3 loop within the Ras-like domain as mapped onto the G α i-GTP structure (PDB: 1CIP)¹³ in Fig. 2D. Of note, several resonances within these key regions are undetectable in the GDP-bound state at all pH values (Fig. 2D, black) which somewhat limits NMR analyses. These findings are consistent with previous work showing that residues associated with the Switch regions in the GDP-bound state of G α i exhibit enhanced backbone dynamics and are broadened and undetectable compared to the resonances associated with the GTP-bound state⁹. Taken together, 2D NMR analyses, CD data and tryptophan fluorescence data indicate that the Switch regions of G α i-GDP are more dynamic and less structured at pH 6.4–6.8 compared to pH 7.2–7.5.

pH-dependence of GDP binding to G α i

G α i, when bound to GDP, adopts a conformational ensemble and dynamic properties distinct from that of the GTP-bound state⁹. This in turn drives recognition of regulatory factors and downstream targets. Given our findings that pH modulates G α i-GDP structure, stability and dynamics, we asked if pH could alter GDP binding to G α i protein. For that purpose, we performed Mant-GDP dissociation assays. For these assays, the rate of GDP dissociation was determined by adding excess GDP to Mant-GDP loaded G α i and monitoring Mant fluorescence changes (by FRET upon tryptophan excitation) as a function of time at different pH values (Fig. 3A–B). As shown in Fig. 3C–D, small differences in GDP dissociation rates were observed over the pH range (pH 6.8–7.4), indicating that GDP binding is not significantly altered at

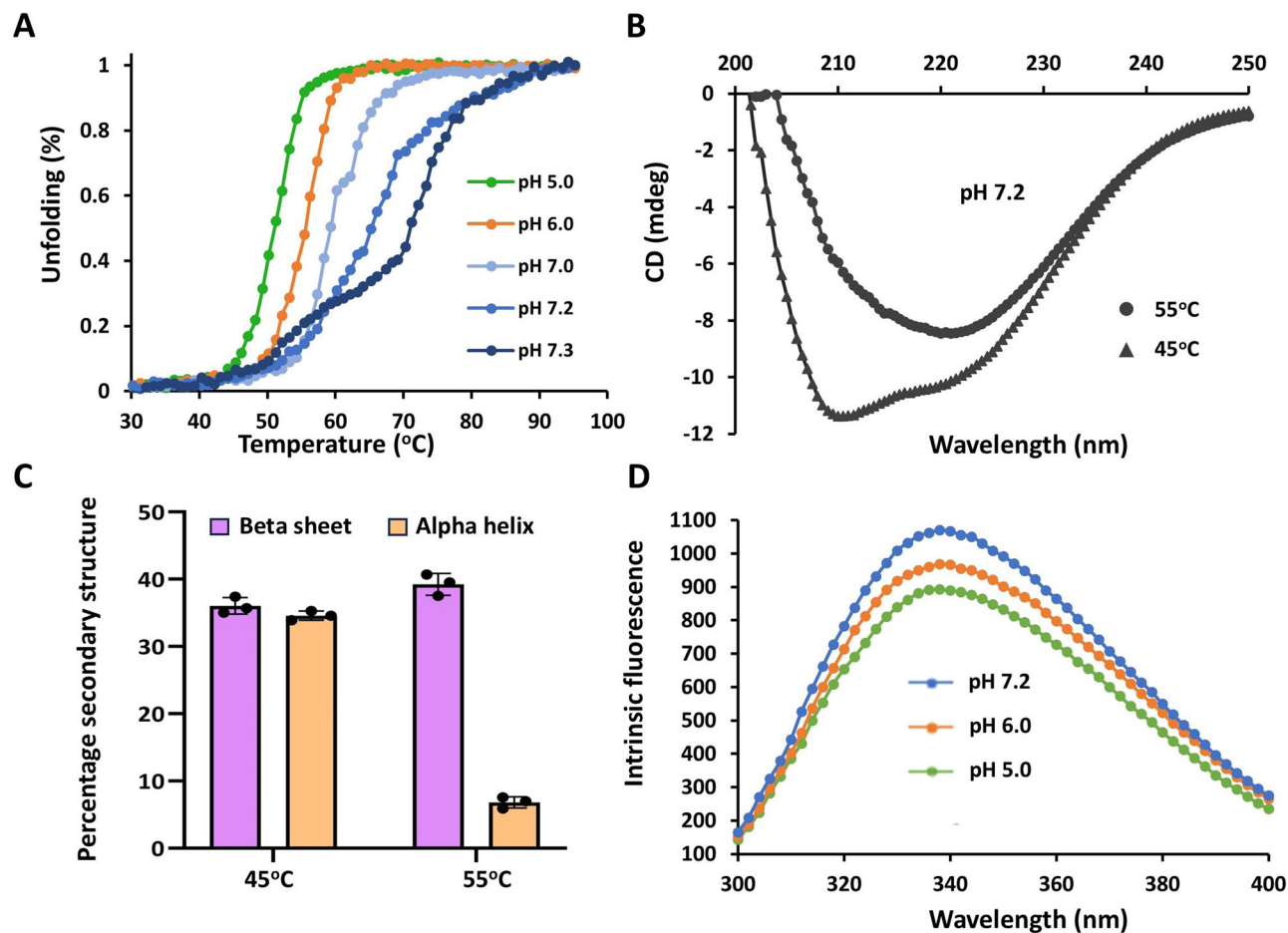


Fig. 1 | Gαi-GDP thermostability and intrinsic tryptophan fluorescence are pH dependent. **A** Representative CD thermal melt (222 nm, 30–95 °C) of Gαi-GDP (15 μM) demonstrates enhanced thermostability at higher pH ($N=3$).

B Representative CD spectral scans (200–250 nm) of Gαi-GDP collected at 45 and 55 °C and pH 7.2 ($N=3$). **C** Bar graph derived from temperature-dependent CD

spectral scans shows a loss of alpha-helical but not beta-sheet secondary structure. Data are averages of $N=3$ independent experiments (\pm SE). **D** Representative intrinsic tryptophan fluorescence spectra of Gαi-GDP (2 μM, excitation = 280 nm, emission = 300–400 nm) show enhanced fluorescence as the pH is increased from 5 to 7.2 ($N=2$ independent experiments, $n=6$ total replicates).

physiological pH. Interestingly, enhanced rates of GDP dissociation were observed at pH values below the cytosolic physiological pH regime.

Identification of pH-sensing residues of Gαi-GDP

To elucidate the molecular basis for pH-dependent stability and conformational dynamic changes in Gαi-GDP, we sought to identify key residues involved in pH sensing. As extensive broadening of resonances in GDP-bound spectra prevented pK_a determination by NMR, we examined and identified two networks of charged residues in regions, designated as the ‘GDP release network’ (Fig. 4A) and ‘Switch network’, that showed pH-dependent changes in the NMR spectra (Fig. 2D). The GDP release network contains residues within $\alpha 1$, $\alpha 5$, $\beta 2$ and $\beta 3$, and was previously shown to be important for GDP release during the GPCR-mediated activation of the Gαi¹⁴. We postulated that if this network plays a key role in pH sensing, mutation of charged residues within this network (e.g., H57, H188, K192, D193, H195 and D337) would reduce pH-dependent Gαi thermostability due to protonation and loss of electrostatic interactions that destabilize Gαi tertiary structure.

To examine whether this network modulates pH-dependent changes in stability and nucleotide binding, we generated several Gαi variants and conducted pH-dependent CD thermal melt and nucleotide dissociation assays. To select neutral or uncharged amino acid substitutions that retain or have a minimal effect on Gαi structure,

we employed the Rosetta modeling suite. Rosetta replaces a desired amino acid within the protein with all possible amino acid substitutions and predicts associated free energy changes¹⁵. Substitutions that minimally perturb free energy are predicted to retain protein structure. Using this strategy, we identified four variants (H57T, H188V, K192Q and H195N) predicted to retain Gαi structure and stability. Further, we performed a GDP dissociation assay to evaluate whether the Gαi variants alter nucleotide binding. Both H57T and K192Q variants showed pH-dependent GDP dissociation rates similar to WT Gαi (Fig. 4B). Also, to evaluate pH-dependent thermostability associated with these Gαi GDP release network variants, we acquired CD thermal melts at pH 6 and 7.2. Of note, all four variants retained pH-dependent thermostability similar to WT Gαi (Fig. 4C–E and Supplementary Fig. 2A–C). These findings indicate that the GDP release network does not significantly modulate pH-dependent stability or nucleotide binding.

Given our observations that several charged residues in the Switch regions form stabilizing electrostatic interactions in the active Gαi-GTP bound state, we postulated that the decreased stability of Gαi-GDP at lower pH may result from protonation of pH-dependent Switch network (SW-I, SW-II, SW-III and $\alpha 3$) residues. To test this hypothesis, we selected residues from the ‘Switch network’ shown or predicted to be important for Switch stability in the GTP-bound form of Gαi¹⁶. Then, based on Rosetta prediction, we mutated a subset of charged residues within this network (e.g. R205N, R208Q, E236L, D237G, R242Q and

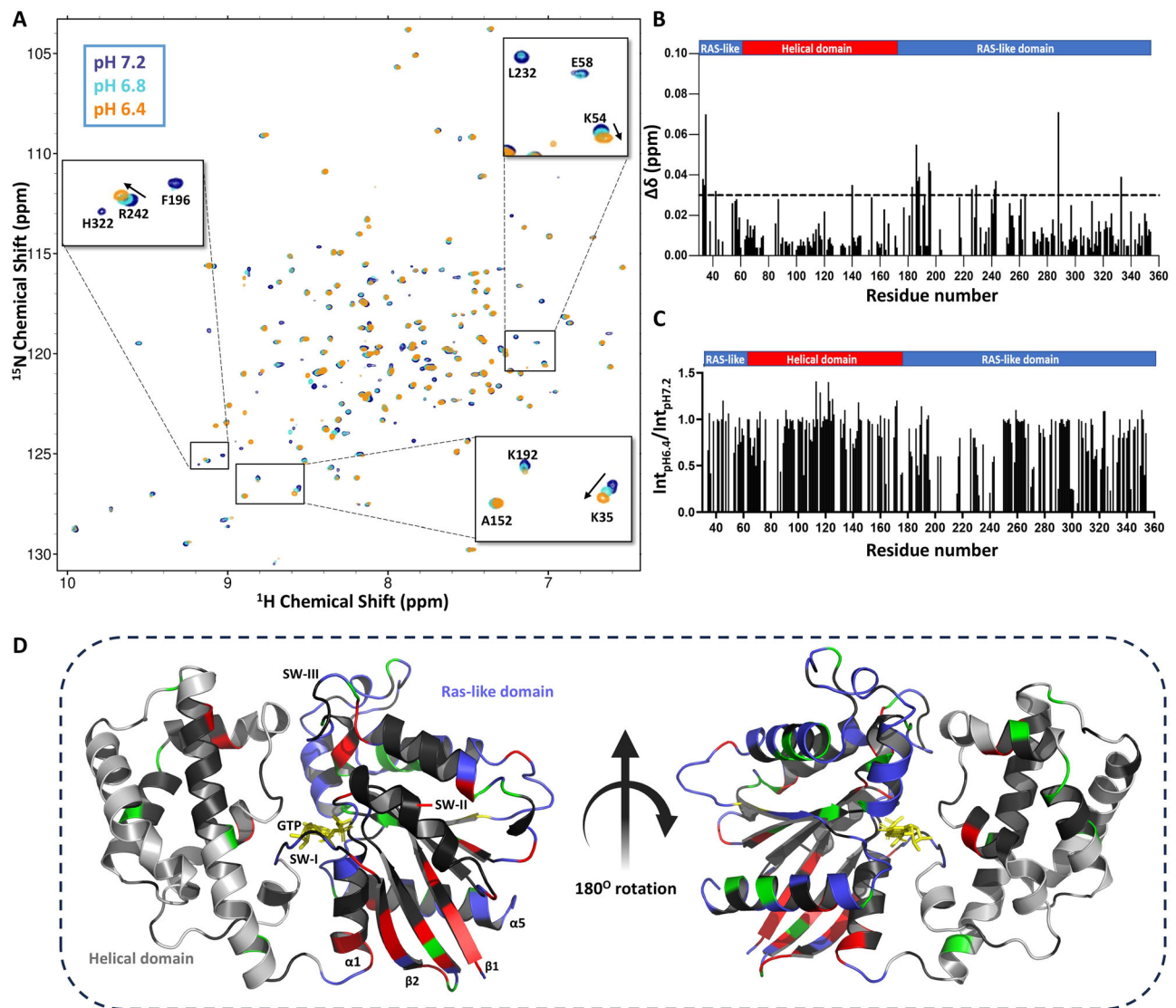


Fig. 2 | Gαi-GDP ^1H - ^{15}N 2D NMR spectral changes as a function of pH.

A Representative 2D ^1H - ^{15}N TROSY-HSQC spectral overlay of ^2H , ^{13}C , ^{15}N -enriched Gαi-GDP (230 μM) at pH 6.4, 6.8 and 7.2, highlighting peak shifts and line broadening. Spectra were acquired on a Bruker Avance III 850 MHz instrument at 25 $^{\circ}\text{C}$ ($N = 2$). **B** Chemical shift perturbation (CSP) and **(C)** peak intensity ratio ($\text{Int}_{\text{pH } 6.4} / \text{Int}_{\text{pH } 7.2}$) for pH 6.4 and 7.2. The striped horizontal line represents the mean value of $\Delta\delta$ amide. Most of the residues that show significant CSP, and broadening lie within

the Ras-like domain ($\alpha 1$, $\alpha 5$, $\beta 1$, $\beta 2$ and $\beta 2$ - $\beta 3$ loop). **D** Spectral differences are highlighted on a ribbon diagram of GTP-bound Gαi (PDB: 1CIP). NH residues with CSP greater than 0.03 ppm are represented in red. Residues with decreased peak intensity (line broadening) at pH 6.4 relative to pH 7.2 are shown in green. Residues missing or unassigned are shown in black color, while unaffected residues are shown in blue and gray color for the Ras-like domain and helical domain respectively.

E245N) predicted to least perturb Gαi structure (Fig. 5A). As shown in Fig. 5B-E, two of the variants located in SW-III (E236L and D237G) and $\alpha 3$ (E245N), respectively, showed a reduction in pH-dependent thermostability change between pH 6 and 7.2 relative to WT Gαi. Moreover, a ‘double variant’ consisting of two substitutions (E236L + E237G) from SW-III showed further reduction in pH-induced stability but not complete abolishment of pH dependence (Fig. 5F). Yet, a ‘triple variant’ containing all three substitutions (E236L + D237G + E245N) effectively eliminated pH-induced stability changes relative to WT Gαi (Fig. 5G), suggesting a key role for these three Switch network residues (E236, D237 and E245) in forming a ‘pH-sensing network’ and stabilizing Gαi at higher pH. To further confirm the role of these residues in the pH-sensitive network, we generated a compensatory variant of Gαi (E236D + D237E). Notably, this Gαi-GDP compensatory variant retains pH-dependent thermostability changes (Supplementary Fig. 3) similar to WT Gαi-GDP. We also monitored tryptophan fluorescence of the

GDP-bound Gαi ‘triple variant’ as a function of pH to examine changes in solvent accessibility of SW-II residue W211. Consistent with the loss in pH-dependent thermal stability, we found that the ‘double variant’ and ‘triple variant’ reduce and abolish pH-dependent fluorescence intensity changes relative to WT Gαi-GDP, respectively (Fig. 5H and Supplementary Fig. 4). As the CD thermal and fluorescence profiles associated with the ‘double variant’ and ‘triple variant’ mimic WT Gαi at lower pH (pH 6), we refer to these variants as ‘low pH mimetics’. To further confirm that the identified residues participate in pH-sensing, we performed NMR analyses for the ‘double variant’ at pH 6.4 and pH 7.2 in comparison to WT Gαi-GDP. As shown in Supplementary Fig. 5, pH-dependent changes in peak intensity and chemical shift perturbations are significantly reduced in the ‘double variant’ with respect to WT Gαi. Taken together, our results point to key residues in the Switch network that regulate pH-dependent stability and conformational dynamic properties.

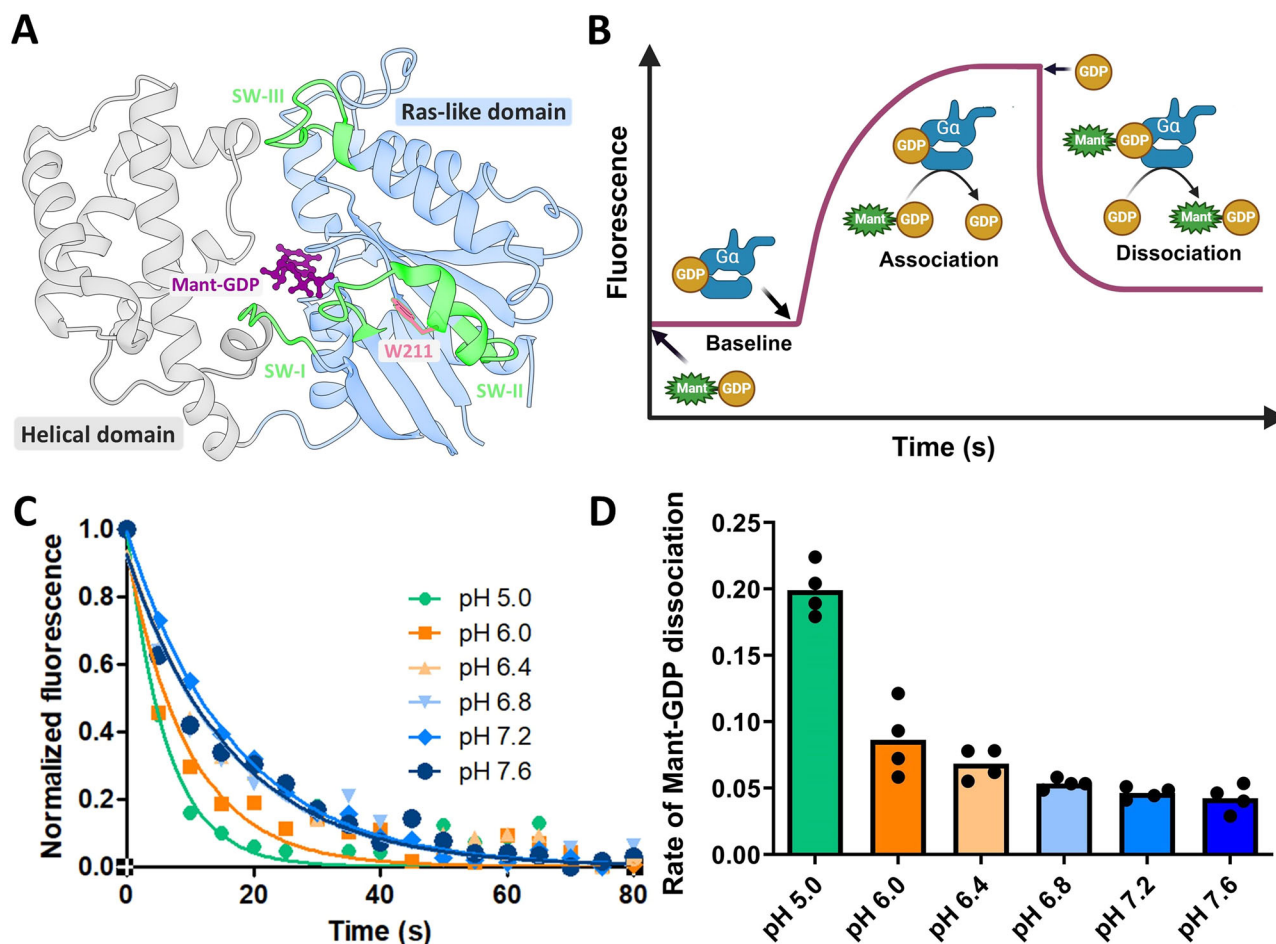


Fig. 3 | Rates of G α i GDP nucleotide dissociation vary with pH. **A** Structure of G α i (PDB: 1GP2) highlighting the position of tryptophan (W211) in Switch II and bound Mant-GDP. **B** Diagram displaying experimental setup of FRET-based determination of G α i nucleotide association and dissociation rates. **C** Rate of GDP dissociation from G α i as a function of pH by monitoring the time-dependent decrease in FRET

emission of G α i-loaded Mant-GDP at 445 nm upon the addition of 7.5 μ M GDP. **D** The rate of Mant-GDP dissociation decreases as the pH increases. Data are averages of $N = 2$ independent experiments with $n = 4$ total replicates. Figure 3B was created in BioRender. Prakash, A. (2025) <https://BioRender.com/b57r381>.

MD simulations of pH-dependent G α i-GDP conformation

We identified three residues in G α i-GDP, including two SW-III residues (E236, D237) and one α 3 residue (E245), that appear to play a key role in pH-dependent stability and conformational dynamics. To evaluate how these residues form pH-dependent electrostatic interactions that stabilize the Switch regions at higher physiological pH, we employed molecular dynamics (MD) simulations. We generated G α i structures using the G α i-GDP crystal structure (PDB: 1GP2) as a starting point and then changed the protonation state of side chains associated with residues E236, D237 and E245 to simulate a low pH and high pH state, respectively.

Analysis of MD simulation trajectories of G α i-GDP collected for 1000 ns in ‘charged’ versus ‘uncharged’ states revealed higher root mean square deviation (RMSD) in the protonated or ‘uncharged’ state. As shown in Fig. 6A, the RMSD of the ‘uncharged’ state indicated higher dynamics, with RMSD values around 3.12 ± 0.41 Å, while the ‘charged’ state exhibited lower RMSD values of 2.42 ± 0.37 Å. Based on analyses of the MD trajectories, we attribute the RMSD reduction associated with the ‘charged’ state to the formation of salt-bridge interactions within the Switch regions. To further probe the dynamic properties of the G α i-GDP Switch regions, we calculated residue-specific root mean square fluctuation (RMSF) in the ‘uncharged’ state (Fig. 6B) and mapped RMSF changes onto the G α i-GTP crystal structure (PDB: 1CIP)¹³ (Fig. 6C). As shown in Figs. 6B and C, our simulations suggest increased dynamics of SW-I, SW-II and SW-III in the ‘uncharged’ state compared

to the ‘charged’ state, consistent with reduced thermal stability at lower pH. Overall, these findings suggest that at lower pH (pH 6.4–7), there is a loss of electrostatic interactions within the Switch regions of G α i-GDP that promotes enhanced dynamics, consistent with observations from NMR and CD data.

Representative snapshots extracted from the MD trajectories of G α i-GDP provide insights into the dynamic behavior of G α i as a function of pH. In the ‘charged’ or higher pH state of G α i-GDP, we observe the transient formation of three critical salt-bridge interactions: E236 interacting with R205, E245 with R208 and E245 with K248 (top panel of Fig. 6D). Notably, these residues form electrostatic interactions as observed in crystal structures of active G α i-GTP state (PDB: 1CIP)¹³ and play a pivotal role in the stabilization of the Switch regions. In particular, E236 and D237 side chains from SW-III interact with R205 and R208 in SW-II, whereas α 3 residue E245 interacts with R208¹⁶. In the ‘uncharged’ state of G α i-GDP (lower pH), conformations extracted from MD simulation trajectories exhibit a notable absence of these salt-bridge interactions (bottom panel of Fig. 6D and Supplementary Table 1), which we attribute to the destabilization of the SW-III and SW-II region (Fig. 6E). As a result, the protein displays increased dynamics, as evidenced by higher RMSD and RMSF values, suggesting greater structural fluctuations. The protonation of key residues interferes with the formation of electrostatic interactions, leading to their destabilization. Consequently, the protein becomes more dynamic and less thermally stable under more acidic conditions. The 2D and 3D Gibbs

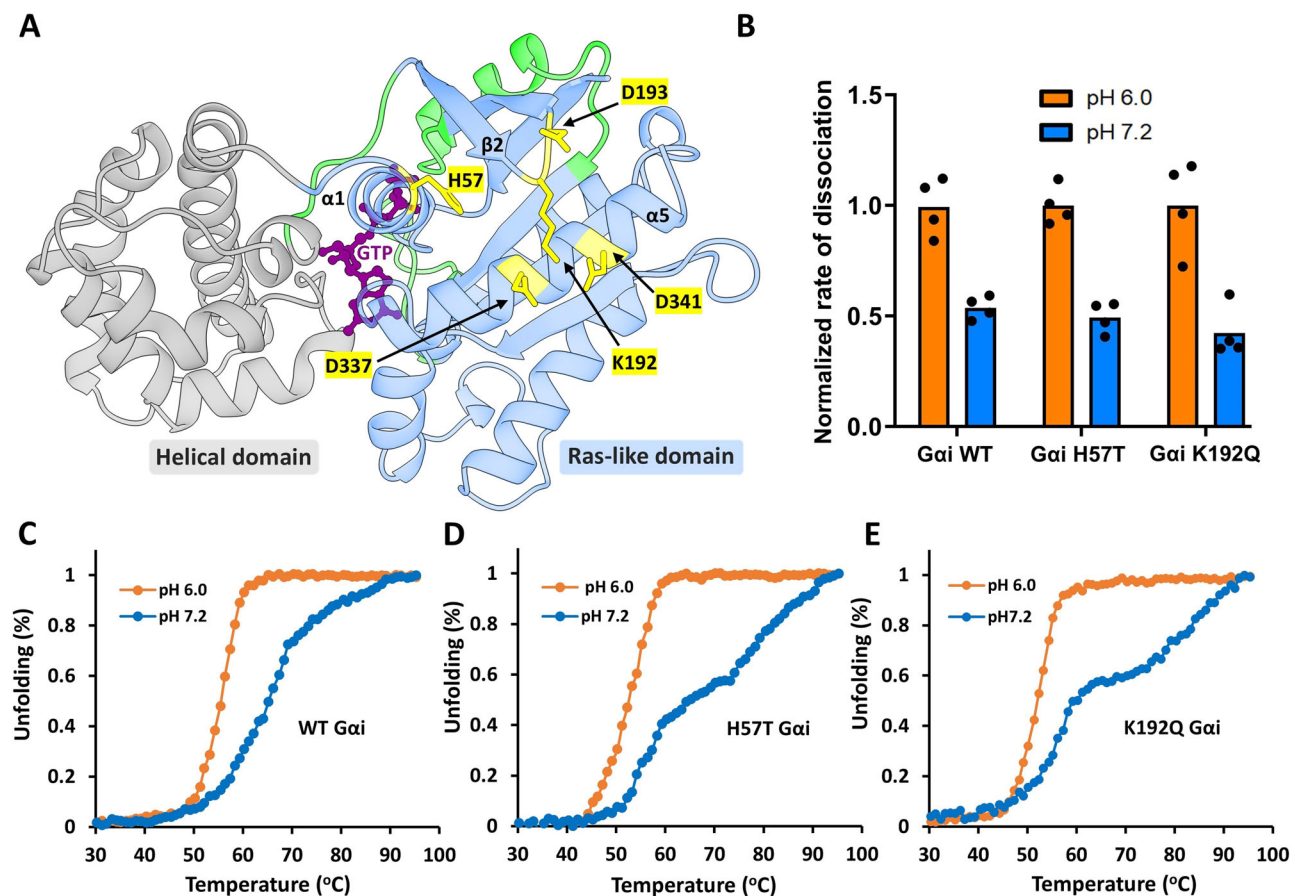


Fig. 4 | pH-dependent nucleotide exchange and stability of Gαi variants from the GDP release network. **A** Ribbon diagram of GTP-bound Gαi (PDB: 1CIP) highlighting electrostatic network near GDP release network. The helical domain, Ras-like domain and the Switch regions are shown in gray, blue and green, respectively. Charged residues are shown as yellow sticks. **B** Rates of GDP

dissociation are compared for WT Gαi and Gαi variants (H57T and K192Q) as a function of pH. Data are averages of $N = 2$ independent experiments with $n = 4$ total replicates. **C** Representative CD melt profile of GDP-bound WT and variant Gαi ($N = 2$), Gαi H57T (**D**) and K192Q (**E**) proteins. Gαi release network variants retain pH-dependent thermal unfolding profiles similar to that of WT Gαi.

Free Energy plots illustrate the distribution of Gαi conformational states in both the ‘charged’ and ‘uncharged’ states. In the ‘charged’ state, the global minimum energy conformation shows the presence of key salt-bridge interactions, specifically E236-R205, E245-R208 and E245-K248, which are absent in the ‘uncharged’ state, reflecting a different conformational energy profile (Fig. 6E).

To predict how a pH-dependent disorder-to-order transition in Gαi-GDP affects the interaction of Gαi with binding partners, we analyzed interactions with GPCRs or Gβγ. Agonist-simulated GPCRs interact with Gαi-GDP through α5, which is part of the GDP release network. As inspection of the crystal structure of β1-adrenergic receptor with Gαi and Gβγ (PDB: 7SOF)¹⁷ indicates that GPCR engagement does not alter the structure of SW-II and SW-III¹⁷, we predict that pH changes over the physiological pH range do not significantly modulate GPCR interactions with Gαi. Our NMR data indicates that key residues within α5 (e.g., F336, T340, I343, K345, etc.), which directly interact with GPCRs¹⁷, lack pH-dependent chemical shifts or peak intensity changes, suggesting that the interaction of Gαi with GPCRs is not altered over the 6.4–7.2 pH range (Fig. 2D). In support of these observations, the MD-derived RMSD (Supplementary Fig. 6) and residue-wise RMSF (Fig. 6B) of α5 in ‘charged’ and ‘uncharged’ states of Gαi-GDP are similar, indicating that the conformational dynamic properties of α5 are unaffected by pH changes. Overall, the NMR and computational analyses strongly suggest that pH-dependent structural changes in Gαi do not affect α5 and likely interactions with GPCRs. However, these observations do not entirely rule out the possibility that the interaction between Gαi and GPCRs could be indirectly

influenced by pH. Conversely, Gβγ binds to Gαi-GDP through the SW-II region, which is more dynamic in the GDP-bound form than the GTP-bound form. We propose, consistent with a recent study¹⁶, that residues from the Switch network (including the three identified pH-sensing residues) provide stabilization of SW-II in the GTP-bound form of Gαi, which in turn prevents interaction with Gβγ. Since our MD data show enhanced dynamics in the protonated or lower pH state of Gαi-GDP, we predict that at the lower end of the physiological pH range, Gαi engages Gβγ with higher affinity, resulting in the downregulation of Gαi and Gβγ mediated downstream signaling. To evaluate this hypothesis computationally, MD simulations were performed on the Gαi-Gβγ trimeric complex in both ‘charged’ and ‘uncharged’ states of Gαi. The results show an increased number of H-bonds between Gαi and Gβγ in the ‘uncharged’ state of Gαi with respect to the ‘charged’ state, indicating increased binding between Gαi and Gβγ (Fig. 7A). To directly assess binding interactions between Gαi and Gβγ, we estimated the binding energy (ΔG) of Gαi in ‘charged’ versus ‘uncharged’ states with Gβγ using MM/PBSA analysis. In the ‘uncharged’ state, Gαi exhibits a stronger binding affinity (-74.99 ± 20.04 kcal/mol) compared to its ‘charged’ state (-60.13 ± 12.97 kcal/mol), suggesting that protonation of residues E236, D237 and E245 in Gαi promotes its interaction with Gβγ (Fig. 7B). Residue-wise decomposition of binding energy contributions over time for both the ‘charged’ (Fig. 7C) and ‘uncharged’ states (Fig. 7D) of Gαi further shows enhanced interaction associated with several surface residues between Gαi and Gβγ in the ‘uncharged’ state. Specifically, Figs. 7E and F highlight stronger interactions between Gαi residues E207, K210 and K211 with residues D186

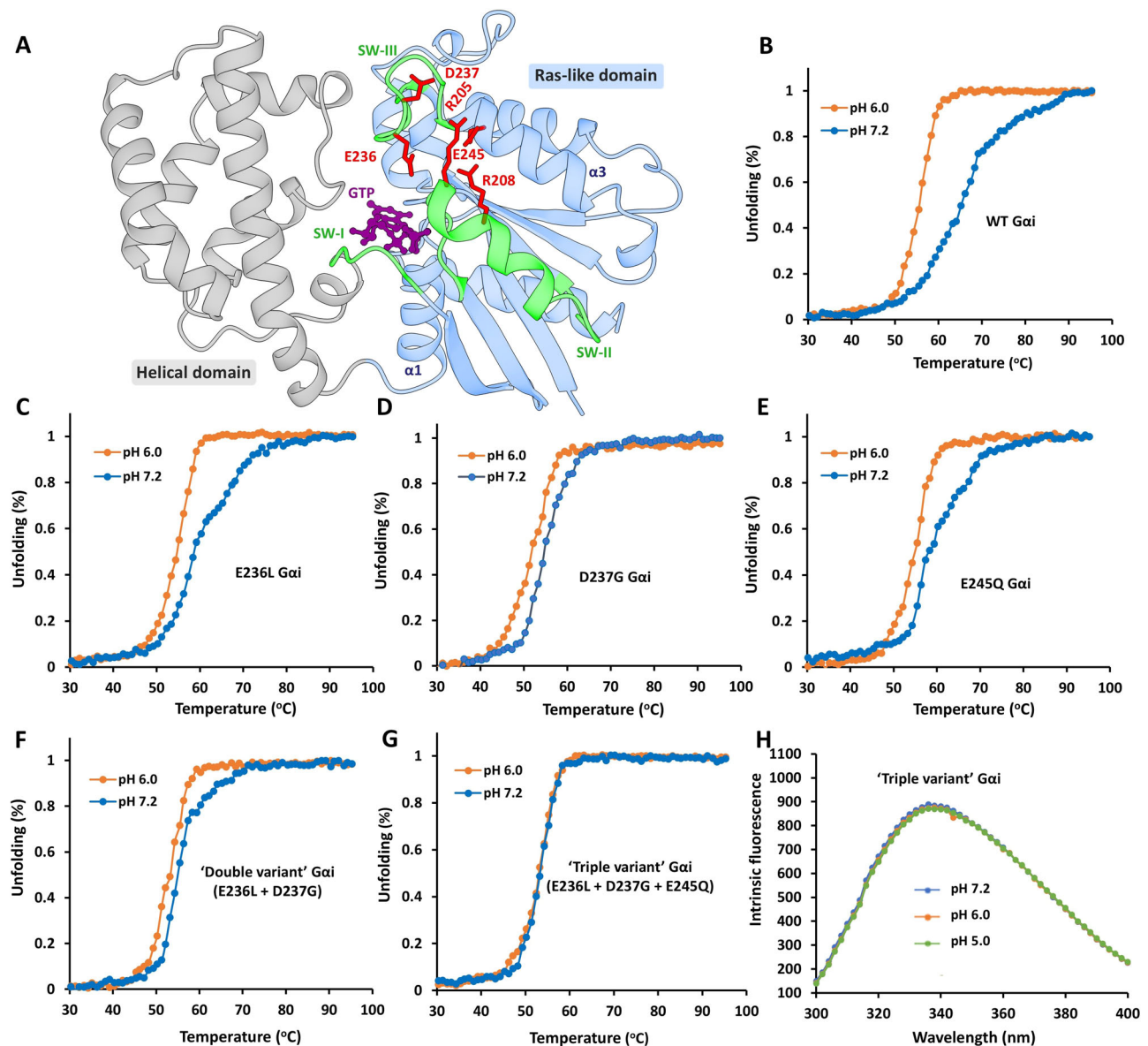


Fig. 5 | pH-dependent stability and tryptophan fluorescence of Gαi Switch network variants. **A** The putative pH-dependent electrostatic network (red) within the Switch regions (green) is highlighted on the ribbon diagram of GTP-bound Gαi (PDB: 1ICIP). **B** Comparison of representative CD melt profiles obtained from two independent experiments ($N = 2$) at pH 6.0 and 7.2 for WT Gαi-GDP, **(C)** Gαi-GDP E236L, **(D)** Gαi-GDP D237G, **(E)** Gαi-GDP E245Q **(F)** Gαi-GDP 'double variant' (E236L + D237G), and **(G)** a Gαi-GDP 'triple variant' (E236L + D237G + E245Q). The CD thermal profile for Gαi single and double variants shows decreased pH-

dependent thermal stability while the 'triple variant' shows a complete loss of pH-dependent thermal stability compared to WT Gαi. **H** Representative intrinsic tryptophan fluorescence spectra of Gαi-GDP 'triple variant' (2 μM, excitation = 280 nm, emission = 300–400 nm) as a function of pH ($N = 2$ independent experiments, $n = 6$ total replicates). Consistent with the CD results, pH-dependent intrinsic tryptophan (W211) fluorescence observed for WT Gαi-GDP is abolished for the Gαi 'triple variant' supporting a role for E236, D237 and E245 in pH-dependent stability and structural changes.

and D228 of Gβγ in the 'uncharged' state compared to the 'charged' state. MD simulations were also performed for the Gαi 'triple variant' in complex with Gβγ. Similar to the 'uncharged' state of Gαi, the 'triple variant' possesses stronger binding to Gβγ (Supplementary Fig. 7). Taken together, our computational analyses suggest that Gαi-Gβγ interactions are modulated by pH, with lower pH_i enhancing Gαi-Gβγ trimeric complex formation.

Analyses of pH-dependent Gαi-Gβγ interactions

Gβγ interacts with the dynamic SW-II region of Gαi in the GDP-bound state. Upon upstream activation by GPCRs, Gαi becomes activated through the exchange of GDP for GTP. In the GTP-bound state, the SW-II region becomes rigid and Gαi is unable to adopt a conformation compatible with binding to Gβγ. Given our findings that Gαi

adopts a more dynamic state at the lower end of the physiological pH range (6.8–7.0), we hypothesized that Gαi-GDP engages Gβγ with a higher affinity at lower pH, whereas higher pH (7–7.5) promotes Gβγ release from Gαi-GDP. Consistent with this hypothesis, our MD analyses indicate that pH-mediated changes in the protonation state of key Switch residues alter electrostatic interactions and the dynamic properties of Switch regions of Gαi in its GDP-bound form which may, in turn, modulate Gαi-Gβγ interactions and thus the downstream signaling. To experimentally evaluate the pH dependence of Gαi-Gβγ interactions in a cellular context, we conducted bioluminescence resonance energy transfer (BRET) assays in human HEK293T cells. For these assays, WT Gαi was tagged with Renilla luciferase (RLuc) and co-transfected with Gγ tagged with GFP2, Gβ and the neurotensin receptor.

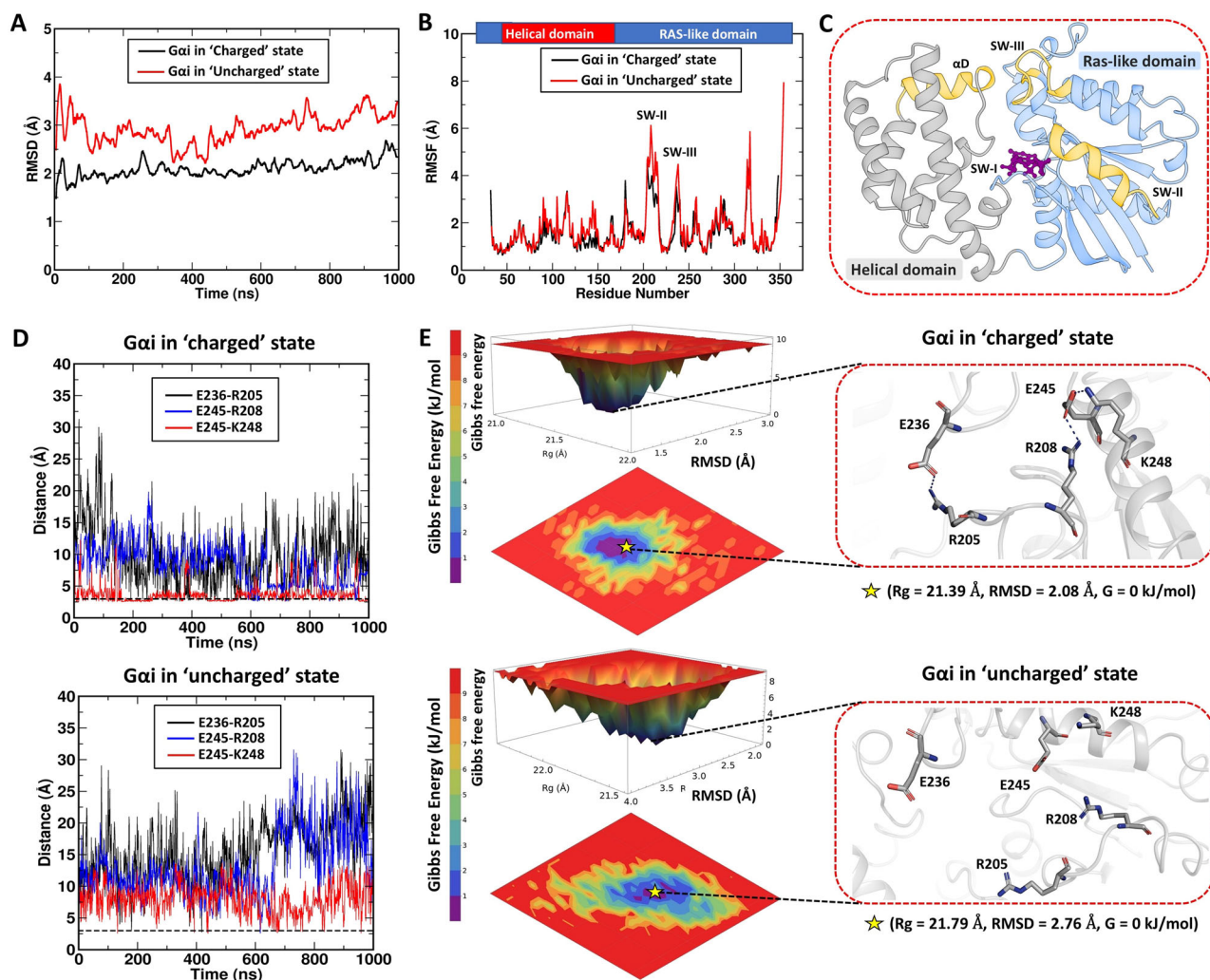


Fig. 6 | MD simulations of WT Gαi-GDP highlighting ‘uncharged’ and ‘charged’ states of E236, D237 and E245. To mimic the low pH state of Gαi-GDP (‘uncharged’ state), E236, D237 and E245 side chains were protonated using the Gαi-GDP crystal structure (PDB: 1GP2) as a starting point. MD simulations of Gαi-GDP in both ‘charged’ (deprotonated) and ‘uncharged’ (protonated) states were performed for 1000 ns in triplicate. **A** Root means square deviation (RMSD) of Gαi Cα atoms in ‘charged’ (black) and ‘uncharged’ (red) states, as determined by superposing each frame of the trajectory to the corresponding starting structure. Gαi-GDP shows higher RMSD in the ‘uncharged’ state compared to the ‘charged’ state, indicative of enhanced dynamics and consistent with NMR line broadening at lower pH. **B** Residue-wise root mean square fluctuation (RMSF) derived from MD trajectory (1000 ns) of ‘charged’ (black) and ‘uncharged’ (red) Gαi-GDP, highlights enhanced Switch dynamics in the ‘uncharged’ state. **C** Mapping of residues that display higher RMSF (yellow) on the Gαi-GTP crystal structure (1CIP) indicates that both SW-II, SW-

III and αD become more dynamic in the ‘uncharged’ state compared to the ‘charged’ state, consistent with the reduced thermal stability and intrinsic tryptophan fluorescence at lower pH. **D** Salt-bridge distances for critical interactions between E236-R205, E245-R208 and E245-K248 in the ‘charged’ (top) and ‘uncharged’ (bottom) states. In the ‘charged’ state, the E245-K248 interaction remains stable throughout the trajectory, while the E245-R208 interaction stabilizes in the final 350 ns. In contrast, none of these interactions induce stable salt-bridges in the ‘uncharged’ state. The 3.2 Å salt-bridge distance cutoff is depicted with a dotted line. **E** Gibbs Free Energy plots for ‘charged’ and ‘uncharged’ states. The 2D and 3D free energy landscapes illustrate the distribution of Gαi conformational states. In the global minimum energy conformation of the ‘charged’ state, salt-bridge interactions between E236-R205, E245-R208 and E245-K248 are present. However, these interactions are absent in the ‘uncharged’ state. The star symbol indicates the position of the minima on the projection.

Upon stimulation with neurotensin (agonist), the Gα-RLuc (energy donor) and Gβγ-GFP2 (energy acceptor) dissociate with receptor-catalyzed dissociation of the Gαi-Gβγ complex measured by comparing energy transfer from donor to acceptor (Fig. 8A). To directly evaluate whether lower intracellular pH (pH_i) reduces Gαi-Gβγ dissociation, we employed two distinct approaches to alter and monitor pH_i changes in HEK293 cells. Intracellular pH was determined by treating HEK293 cells with BCECF dye and determining the ratio of emission intensity (detected at 535 nm) when the dye is excited at ~490 nm and ~440 nm¹⁸. To convert the fluorescence ratios to pH_i, a calibration curve of HEK293 cells treated with 5 μM nigericin was generated (Supplementary Fig. 8A). Treatment of cells with nigericin makes the pH_i equivalent to extracellular pH (pH_e) thus the calibration

curve for pH_e vs. fluorescence can be used to convert fluorescence into cellular pH_i. To alter the pH_i, we compared results using two different strategies. One common method to reduce pH_i is the addition of Carbonyl cyanide 4-(trifluoromethoxy) phenylhydrazone (FCCP), an uncoupler of oxidative phosphorylation in the mitochondria¹⁹. We determined that cells treated with 1 μM FCCP generated a pH_i of -7.04, while at 5 μM the pH_i was further reduced to 6.85 (Supplementary Fig. 8B). As one drawback of FCCP is that it can alter mitochondrial energetics, we employed a second method to alter pH_i by modulating extracellular pH. As shown in Supplementary Fig. 8C, lowering extracellular pH to 6 and 5 caused a reduction in pH_i to 6.9 and 6.7, respectively, which is consistent with a previous report²⁰. To evaluate the pH dependence of Gαi-Gβγ dissociation in cells, we conducted the

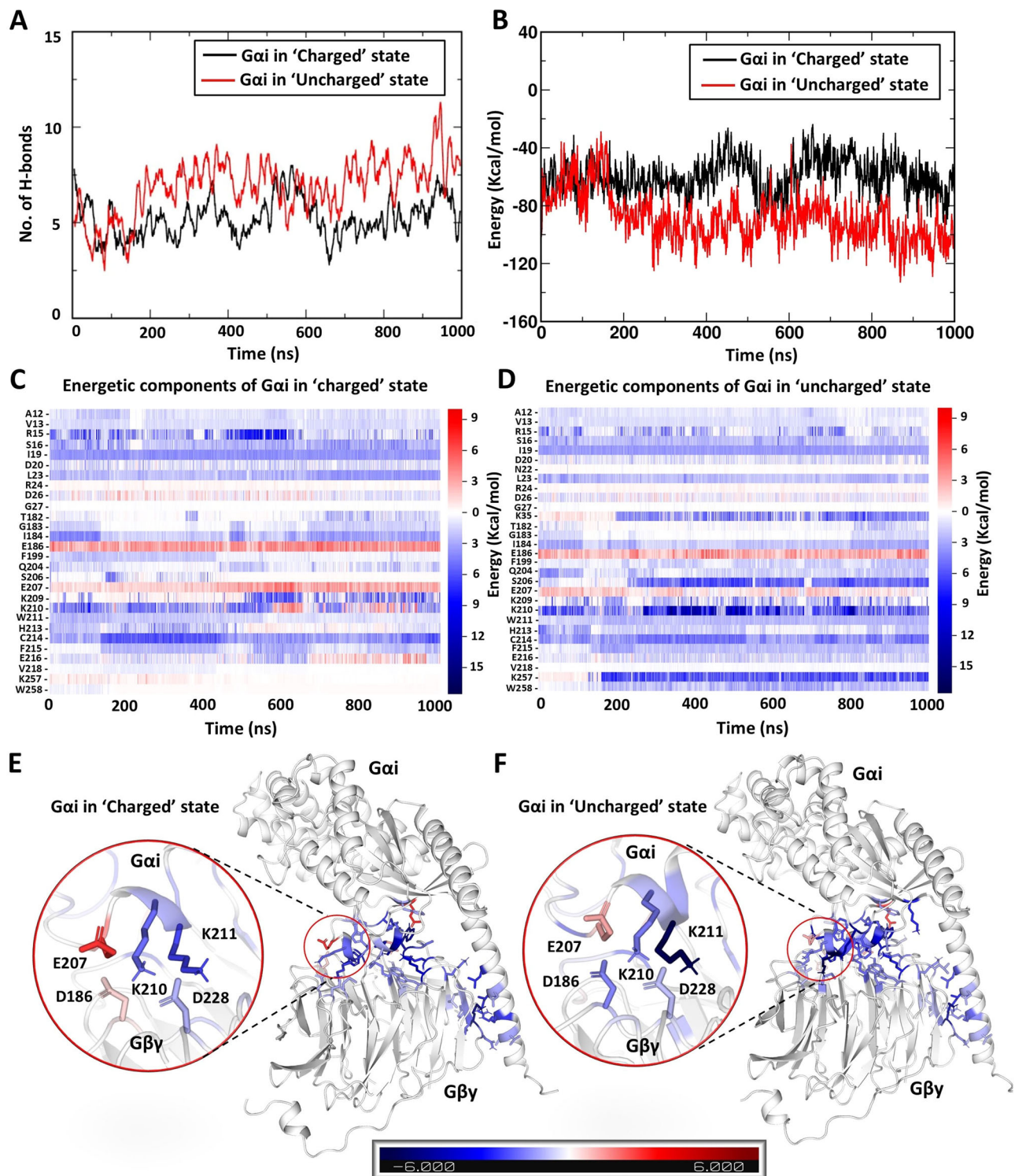


Fig. 7 | MD simulations of ‘uncharged’ and ‘charged’ states of WT Gai-GDP in complex with Gβγ. MD simulations of the Gai-Gβγ complex (PDB: 1GP2) in Gai ‘charged’ and ‘uncharged’ states were performed for 1000 ns. **A** A higher number of H-bonds were formed for the ‘charged’ state (black) of Gai compared to the ‘uncharged’ state (red) suggesting stronger binding affinity of Gai with Gβγ in the ‘uncharged’ state of Gai. **B** Comparison of the binding energy (ΔG) between Gai and Gβγ in ‘charged’ versus ‘uncharged’ states using MM/PBSA analysis. In the ‘uncharged’ state, Gai possesses stronger binding affinity, indicating that protonation of E236, D237 and E245 in Gai enhances its interaction with Gβγ. Residue-

wise decomposition of binding energy contributions of Gai over time for both the ‘charged’ (C) and ‘uncharged’ (D) states of Gai. Structural mapping of the per-residue binding energy decomposition of Gai and Gβγ onto the Gα-Gβγ trimeric complex (PDB: 1GP2) in the ‘charged’ (E) and ‘uncharged’ (F) states of Gai. The color gradient reflects the per-residue energy contribution, where blue indicates stronger, favorable interactions and red portrays weaker or unfavorable contributions. Blue residues contribute favorably to binding, whereas red residues contribute less favorably.

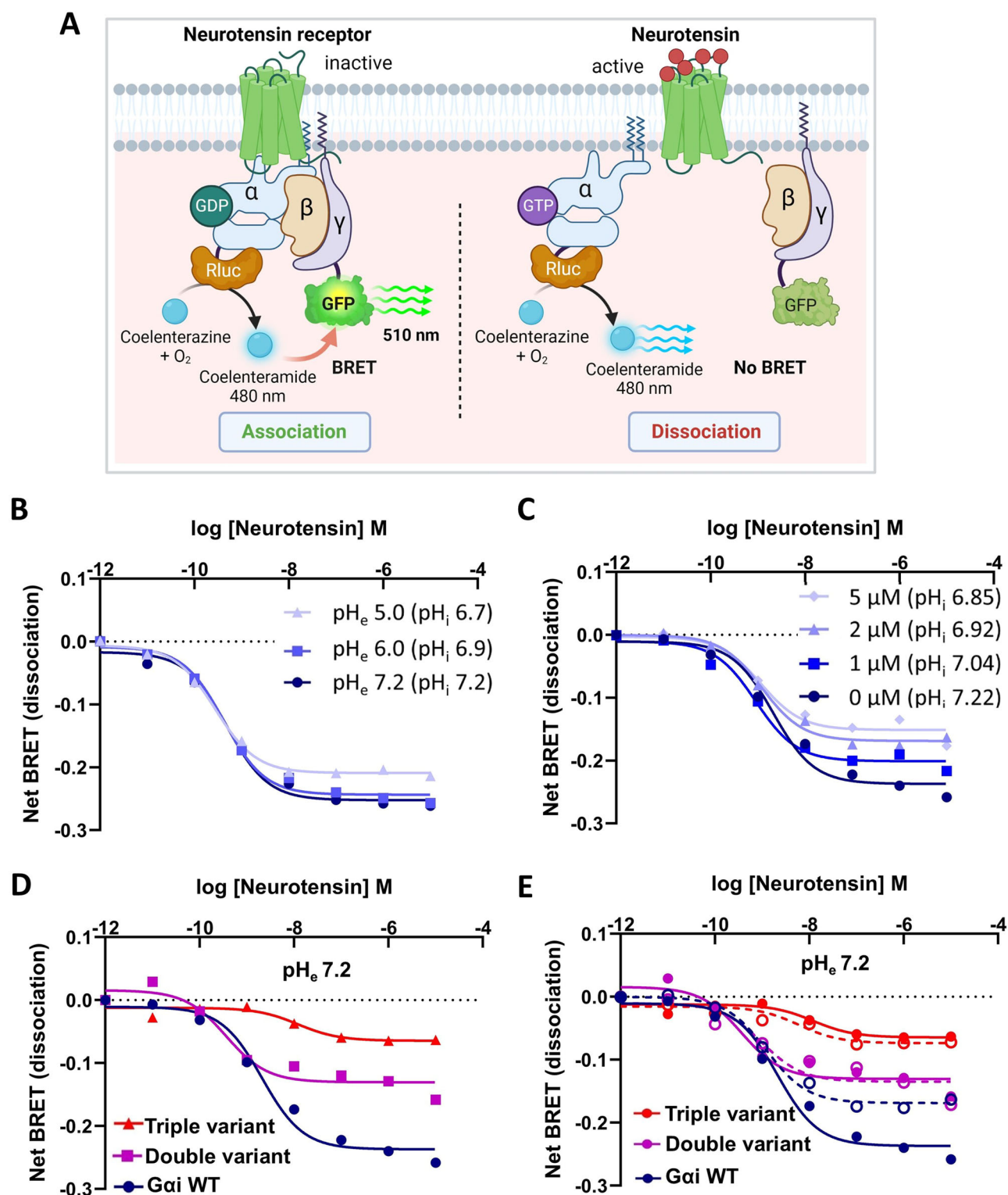


Fig. 8 | BRET assay showing enhanced $G\alpha_i$ - $G\beta\gamma$ association at lower intracellular pH for the $G\alpha_i$ Switch variants. **A Schematic diagram illustrating the BRET assay used to monitor receptor-mediated dissociation of $G\alpha_i$ from $G\beta\gamma$. For this assay, the GPCR neurotensin receptor, $G\alpha$ Renilla luciferase 8 (Rluc8), $G\beta 3$ and $\gamma 9$ -GFP2 fusion constructs were co-transfected in HEK29T cells and fluorescence (515 nm) was monitored after the addition of the substrate and agonist (neurotensin). Net BRET (dissociation) plotted as $\gamma 9$ -GFP2 (acceptor) over $G\alpha$ -Rluc (donor) ratio as a function of log doses of neurotensin (**B**). Neurotensin receptor-mediated BRET shows that lowering pH_i (from 7.2 to 6.7) either by altering**

extracellular pH (**B**) or by the addition of FCCP (**C**), decreases $G\alpha_i$ dissociation from $G\beta\gamma$ in HEK29T cells. **D** WT $G\alpha_i$ shows significantly higher dissociation from $G\beta\gamma$ relative to both 'double variant' and 'triple variant' (low pH mimetics of WT $G\alpha_i$). **E** The difference in net BRET signal as a function of FCCP concentration (0 μ M in solid line and 2 μ M in dotted line) for 'double variant' and 'triple variant' is reduced with respect to WT $G\alpha_i$, confirming key roles for E236, D237 and E245 in pH-mediated $G\beta\gamma$ dissociation for $G\alpha_i$ - $G\beta\gamma$ complex. All BRET data are averages $N=2$ independent experiments with $n=4$ total replicates. Figure 8A was created in BioRender. Prakash, A. (2025) <https://BioRender.com/b57r381>.

BRET assay by expressing WT $\text{G}\alpha_i$ in HEK293 cells and inducing acidosis in the cytoplasm by either altering extracellular pH (Fig. 8B) or by the addition of varying concentrations of FCCP (Fig. 8C). Results obtained from the BRET assays show that lowering pH_i (pH_i 7.2 - 6.85 by FCCP or 7.2 - 6.7 by changing pH_e) reduces agonist-mediated $\text{G}\beta\gamma$ release from $\text{G}\alpha_i$, indicating that $\text{G}\alpha_i$ - $\text{G}\beta\gamma$ interactions are highly dependent on pH over the physiological pH range. Similarly, the basal (agonist unstimulated) BRET shows enhanced $\text{G}\alpha_i$ - $\text{G}\beta\gamma$ association at lower pH_i (Supplementary Fig. 9). Further, to confirm that the observed changes in the BRET signal are due to pH-sensing properties of $\text{G}\alpha_i$ but not the GFP2 (attached to $\text{G}\gamma$), we examined GFP2 fluorescence over a range of intracellular pH values. The results suggest that GFP2 is not pH sensitive within the tested pH range (pH_i 6.7-7.2) (Supplementary Fig. 10).

To further confirm that the modulation of $\text{G}\beta\gamma$ release by pH occurs through the identified 3 residues of $\text{G}\alpha_i$, we performed the BRET assay using the double or triple $\text{G}\alpha_i$ variants. As shown in Fig. 8D, agonist-stimulated $\text{G}\alpha_i$ dissociation from $\text{G}\beta\gamma$ is significantly reduced for the 'double variant' and further reduced for the 'triple variant' with respect to WT $\text{G}\alpha_i$. These variants appear to serve as low pH mimetics, as they produce BRET similar to lower pH_i . This observation was further validated by BRET assays performed for 'double variant' and 'triple variant' at different pH_i (Fig. 8E). Taken together, these findings suggest that enhanced $\text{G}\alpha_i$ - $\text{G}\beta\gamma$ association at lower pH is due to the loss of electrostatic interactions within the Switch network needed for $\text{G}\alpha_i$ - $\text{G}\beta\gamma$ formation.

Discussion

The recognition of pH sensors among biomolecules is crucial in understanding cellular processes, as they play a pivotal role in responding to changes in the proton concentration within the physiological range. While many biomolecules experience alterations in their protonation state, only a specific subset serves as pH sensors, exhibiting pH-dependent functional changes that influence cellular processes.

One of the first pH-sensing proteins to be characterized is hemoglobin. In acidic environments, hemoglobin exhibits a reduced affinity for oxygen, a phenomenon known as the Bohr effect. This effect is due to the protonation of specific amino acid residues, including histidine 146, located in the α subunit of hemoglobin²¹. The protonation of H146 promotes the formation of a salt bridge with a nearby aspartate residue (D94) on the β subunit. This interaction stabilizes the deoxygenated or T-state conformation of hemoglobin, reducing its affinity for oxygen and facilitating the release of oxygen to tissues where it is needed²². After the identification of hemoglobin, proton-sensing ion channels and receptors governing cytosolic pH have been a focal point in research.

More recent structural informatics calculations have shown that buried ionizable networks are a structural hallmark of pH sensitivity³ and the large pK_a shifts exhibited by buried sidechains can be harnessed by proteins to drive pH-dependent changes in structure, stability and function. Indeed, networks of buried ionizable amino acids are conserved in nearly all of the 100+ $\text{G}\alpha$ protein structures in the Protein Data Bank (PDB)³. On that basis, the cell signaling GTPases, $\text{G}\alpha_i$ and its yeast homolog Gpa1, were proposed to function as pH-sensing proteins. The study indicated that the stability of heterotrimeric $\text{G}\alpha$ proteins exhibits a significant dependence on pH levels across a broad range from 5.0 to 8.0. Furthermore, alterations in yeast intracellular pH (pH_i) were shown to promote Gpa1 phosphorylation and subsequently dampen the mitogen-activated protein (MAP) kinase signaling pathway³. In addition to our findings that $\text{G}\alpha_i$ serves as a pH sensor to regulate $\text{G}\alpha$ - $\text{G}\beta\gamma$ interactions, recent studies indicate that select GPCRs possess pH-sensing properties that modulate extracellular signaling to $\text{G}\alpha$ proteins²³⁻²⁵. These and other GPCRs may work independently or synergistically with $\text{G}\alpha_i$ proteins to transduce extracellular pH-dependent signals to changes in intracellular pH^{20,26,27}.

Herein, we find that as pH is raised from 6.8 to 7.3, the thermostability of $\text{G}\alpha_i$ -GDP is greatly enhanced (ΔT_m ~25°C) due to the formation of an electrostatic network within the Ras-like domain. Since thermostability changes for $\text{G}\alpha_i$ -GDP were significantly larger (~20°), relative to the GTP-bound form (~8°), we focused on characterizing pH-dependent structural and dynamic changes associated with the GDP-bound form of $\text{G}\alpha_i$ in this study. Using pH-dependent NMR analyses and thermal stability profiling on WT and mutant proteins, we identified three key pH-sensing residues that drive pH-dependent thermostability changes. Two of the residues (E236 and D237) reside in SW-III of $\text{G}\alpha_i$, while the third residue (E245) lies in the neighboring $\alpha 3$ helix. Mutation of E236 and D237 in SW-III promotes $\text{G}\beta\gamma$ release in HEK293 cells, supporting the role of these residues in pH-dependent electrostatic interactions that allosterically regulate heterotrimer formation. While both Ras and heterotrimeric G-proteins contain SWI and SWII regions, Ras proteins lack the SW-III region found in $\text{G}\alpha$ proteins. The presence of this unique SW-III region in heterotrimeric G-proteins may explain the distinct pH-sensing properties of $\text{G}\alpha_i$, suggesting a potential role of the SW-III region in pH-dependent $\text{G}\beta\gamma$ release from the $\text{G}\alpha_i$ - $\text{G}\beta\gamma$ complex. Of note, the pH-sensing network identified in this study is distinct from the network predicted by Isom et al.³. This computationally predicted pH-sensing network consists of residues at the Ras/helical domain interface (e.g. K46, D150, D200, D229, R242, K270 and K277). However, our NMR and CD data do not support this interaction network, as observable NMR resonances associated with K270 and K277 do not show pH-dependent perturbations and mutation of K46 and R242 did not alter pH-dependent thermostability changes measured by CD.

Large changes in the stability of GTP-bound $\text{G}\alpha_i$ have previously been attributed to the formation of an interaction triad (termed as G-R-E Triad) involving residues G203 and R208 from SW-II and E245 from $\alpha 3$ ¹⁶. This interaction triad is absent from the less stable and more dynamic GDP-bound state^{9,16}. Interestingly, E245 is also one of the residues identified in the pH-sensing network. Moreover, E236 and D237 from SW-III interact with SW-II via residue R205 to stabilize both Switches in the GTP-bound forms of the protein. This nucleotide-dependent disorder-to-order transition is predicted to facilitate the release of $\text{G}\beta\gamma$ from $\text{G}\alpha_i$ resulting in the activation of $\text{G}\alpha_i$. Consistent with these findings, SW-III residues E236 and D237 are involved in pH-dependent electrostatic interactions that appear to allosterically regulate $\text{G}\beta\gamma$ release.

While several pH-sensing proteins contain titratable histidines²⁸⁻³⁰, others such as EmrE contain aspartate and glutamate residues that titrate in the physiological range due to the formation of electrostatic networks^{24,31}. In these systems, even a small reduction in pH can alter the side chain protonation state, leading to the neutralization of charge. In the case of $\text{G}\alpha_i$, we identified a network of three charged residues (E236, D237 and E245) that regulate pH-dependent thermostability changes in the $\text{G}\alpha_i$ -GDP. To investigate how the protonation/deprotonation of these residues alters $\text{G}\alpha_i$ structure and dynamics in the GDP-bound state, we performed MD simulations. As starting points for the simulations, residues E236, D237 and E245 were generated in both 'charged' and 'uncharged' states to represent higher (deprotonated) and lower pH (protonated) states, respectively. Results from these analyses suggest that the protonation of residues E236, D237 and E245 enhances $\text{G}\alpha_i$ -GDP dynamics due to the loss of electrostatic interactions between SW-II, SW-III and $\alpha 3$. Moreover, we find that these residues make transient electrostatic interactions in $\text{G}\alpha_i$ -GDP at higher (>pH 7.2), but not at lower pH. MD simulations performed for the $\text{G}\alpha_i$ - $\text{G}\beta\gamma$ trimeric complex in the 'charged' or 'uncharged' state of WT $\text{G}\alpha_i$ -GDP showed increased binding energy with $\text{G}\beta\gamma$ in the 'uncharged' state, suggesting stronger binding. Based on these MD simulation data, we propose that lower physiological pH promotes $\text{G}\beta\gamma$ binding to $\text{G}\alpha_i$ resulting in attenuation of both $\text{G}\alpha_i$ and $\text{G}\beta\gamma$ mediated signaling pathways (Fig. 9). Consistent with this premise, our findings indicate that at the lower end of the intracellular pH_i range,

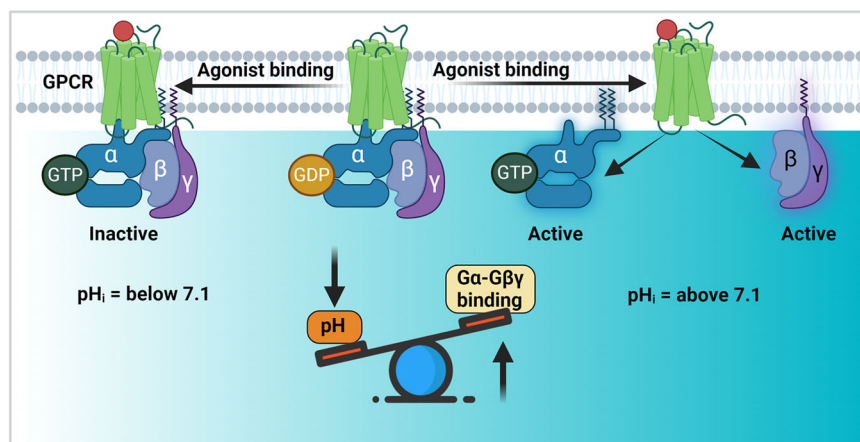


Fig. 9 | Proposed mechanism of Gαi pH regulation. Activation of Gαi by GPCRs leads to the release of Gαi from Gβγ and stimulation of signaling pathways. A distinct mechanism of signaling regulation may occur by intracellular pH regulation. Gαi proteins may act as intracellular pH sensors to regulate Gαi-Gβγ

interactions with lower pH_i, enhancing Gαi-Gβγ interactions to inhibit Gαi signaling. This figure was created in BioRender. Prakash, A. (2025) <https://BioRender.com/b57r381>.

agonist-mediated Gβγ release from Gαi is attenuated due to the loss of pH-dependent electrostatic networks involving key residues within SW-III and α3. This, in turn, is expected to downregulate Gαi and Gβγ downstream signaling.

While the current investigation is centered on the Gαi isoform, sequence and structural alignment with other Gα isoforms (such as Gαs, Gαo and Gαq) reveals conservation within the core pH-sensing network (Supplementary Fig. 11)³. This analysis suggests the potential for pH-sensing properties in other Gα isoforms. The development of pH-insensitive forms of Gαi will facilitate investigations of hormone and neurotransmitter signaling during physiological stresses, such as occur during glucose or oxygen deprivation, leading to changes in cellular pH³².

Methods

Plasmid constructs

A bacterial pET-SUMO vector containing the human Gαi1 gene (GenBank accession no. BC026326) with the first 31 amino acids deleted, N-terminal 6xHis, and SUMO-tag were employed for in vitro analyses. Full-length human Gα containing Renilla luciferase (Gα-RLuc8), Gβ and Gγ containing green fluorescence protein 2 (Gγ-GFP2), and neurotensin receptor (NTR) constructs were employed for cell-based BRET assays^{33,34}.

Site-directed mutagenesis

Gαi variants were generated using the Q5 Site-Directed Mutagenesis kit (NEB). Polymerase chain reaction (PCR) primers were designed using the NEBaseChanger (<https://nebasechanger.neb.com>), an online mutagenesis primer design tool powered by New England Biolabs from Eton Bioscience Inc. Mutagenesis was performed as described³³. The variant constructs were sequenced by Sanger Sequencing (Genewiz) to confirm successful mutagenesis. Sequencing results for Gαi variants were aligned with published sequences of wild-type (WT) Gαi using Clustal Omega (European Molecular Biology Laboratory, Cambridgeshire, UK).

Expression and purification of WT Gαi and its variants

Gαi proteins were overexpressed in *E. coli* Agilent BL21-codon plus (DE3)-RP-X competent cells. Cells were grown at 37 °C for 3–4 hours to achieve an optical density at 600 nm (OD₆₀₀) of 0.60 and then induced with 500 μM isopropylthio-β-galactoside (IPTG). The temperature was then reduced to 18 °C and the cultures were left to grow overnight. After 20 hours, bacteria cells were harvested by centrifugation at 10,000 × g and then resuspended in 50 mL Dialysis buffer

(10 mM KH₂PO₄, 20 mM K₂HPO₄, 0.15 M KCl, 1 mM MgCl₂, 10 μM GDP, 1 mM β-mercaptoethanol (BME), pH 7.0). Resuspended cells were lysed using a sonicator (Fisher Scientific, #CL-334) with an amplitude of 80 and 5 seconds on/off time ratio for 10 mins/L cells. Clarified lysate was passed through the Ni-NTA column equilibrated with 1 column volume (CV) (50 ml) of Dialysis buffer, washed with Wash buffer (10 mM KH₂PO₄, 20 mM K₂HPO₄, 0.15 M KCl, 1 mM MgCl₂, 10 μM GDP, 1 mM BME, 30 mM imidazole, pH 7.0) and eluted using the Elution buffer (10 mM KH₂PO₄, 20 mM K₂HPO₄, 0.15 M KCl, 5 mM MgCl₂, 10 μM GDP, 1 mM BME, 250 mM imidazole, pH 7.0). The elution fraction was dialyzed overnight in the presence of a ubiquitin-like-specific protease 1 (ULP1) to cleave the SUMO tag on Gαi. After dialysis, the protein was again passed over a Ni-NTA column equilibrated with 3 CVs of water and 1 CV dialysis buffer. The protein flow-through was further purified by gel filtration chromatography using a Fast Protein Liquid Chromatography (FPLC) system (Äkta Primeplus) and Superdex 75 column (Cytiva). The purity of the protein was verified by SDS PAGE gel electrophoresis, and the concentration of the purified protein was estimated by Thermo Scientific NanoDrop.

Circular dichroism (CD) analyses

Gαi proteins were exchanged in CD buffer (10 mM K₂HPO₄, 500 μM MgSO₄, 500 μM tris(2-carboxyethyl)phosphine (TCEP)) using an Amicon centrifugal filter unit (MilliporeSigma, #UFC901024) at 3800 × g, diluted to 15 μM (pH 6.0, 6.4, 6.8, 7.2, or 7.6) and centrifuged (16,000 × g) for 10 mins at 4 °C. Temperature dependent CD experiments were performed on a Jasco J815CD spectrometer using 15 mM WT or Gαi variants protein in a 1 mm path-length quartz cuvette (Hellma Analytics). Thermal melts were obtained at 222 nm, over a temperature range of 20–95 °C, using a temperature increment of 1 °C/min. The CD spectral scans were collected for Gαi proteins at different fixed temperatures (e.g. 45 °C, 55 °C and 65 °C) by taking CD measurements every 1 nm from 200–250 nm, and secondary structure evaluated using the online server BeStSel (Beta Structure Selection)³⁵.

Intrinsic tryptophan fluorescence

Intrinsic tryptophan fluorescence assays were conducted using 2 mM purified WT or variant GDP-loaded Gαi proteins in 200 ml of assay buffer (20 mM HEPES, 50 mM NaCl, 5 mM MgCl₂, 2 mM DTT) at different pH (pH 5, 6 and 7.2) in the 96 well plate. Gαi proteins were excited at 290 nm and intrinsic tryptophan fluorescence was measured from 300 to 400 nm wavelength using a SpectraMax M4 Series Microplate Reader.

2D NMR experiments

Triple labeled $^2\text{H}/^{13}\text{C}/^{15}\text{N}$ G α i samples were generated by expressing G α i proteins in BL21(DE3)-RIPL *E. coli* cells and growing in minimal medium supplemented with 1 g/L $^{15}\text{NH}_4\text{Cl}$, 3 g/L ^{13}C -glucose, 99% D_2O (Cambridge Isotope Labs)⁹, and purified as described above. Triple labeled G α i was prepared in 20 mM potassium phosphate (pH=7.0), 50 mM potassium chloride, 5 mM DTT (dithiothreitol), 5 mM magnesium chloride, and 50 μM GDP containing 5% (v/v) D_2O . For ^{15}N -enriched WT and G α i variants, bacterially expressed proteins were grown in minimal media containing 1 g/L $^{15}\text{NH}_4\text{Cl}$ as the sole nitrogen source and purified as described above.

2D NMR ^1H - ^{15}N HSQC spectra of G α i were acquired on a Bruker Avance 850 MHz (14.1 T field strength) NMR spectrometer at 25 $^\circ\text{C}$, with a cryogenic (TCI) 5 mm triple resonance probe equipped with a z-axis gradient. The ^1H - ^{15}N HSQC 2D NMR experiments were recorded with 2048 and 224 complex points, 16 and 36 ppm spectral width, 75.36 and 36.10 ms acquisition times in the direct and indirect dimensions, respectively, 16 scans per increment and a recovery delay of 1.25 s. The ^1H - ^{15}N HSQC spectra of G α i-GDP were assigned using a combination of triple resonance experiments, including 3D HNCA, HN(CO)CA, HN(CA)CO and HNCO⁹. TROSY-based pulse sequences were used for sensitivity enhancement. Bruker-TopSpin 4.1.1 was used to process the NMR data and NMRfAM-SPARKY was used to visualize and analyze the NMR spectra³⁶. For assignments, BMRB 30078 was used as a reference spectrum for G α i-GDP³⁷. pH titration studies were performed by calculating the chemical shift perturbation (CSP) of G α i-GDP over a pH range from 6.4, 6.8, 7.0, 7.2, 7.4 and 7.6. Average ^1H - ^{15}N CSP were determined using the formula $\Delta\delta = [(\Delta^1\text{HN})^2 + (\Delta^{15}\text{N}/5)^2]^{0.5}$. PyMOL³⁸ (<https://pymol.org/2/>) and UCSF ChimeraX³⁹ (<https://www.cgl.ucsf.edu/chimera/>) was used to generate all images of molecular structures.

Guanine nucleotide dissociation assay

G α i proteins were exchanged into nucleotide association buffer (20 mM HEPES, 50 mM NaCl, 5 mM MgCl_2 , 2 mM DTT) using an amicon concentrator (3800 g, 15 min, 3 rounds). In the cuvette, 0.75 μM (2'- or 3')-O-(N-Methylanthraniloyl) Guanosine 5'-Diphosphate, Disodium Salt (Mant-GDP, purchased from ThermoFisher) was added to 1 ml of association buffer for the assay. The intrinsic tryptophan (W211) was excited at 280 nm and the Mant-GDP fluorescence intensity at 425 nm was measured as a function of time using a PerkinElmer LS55 luminescence spectrometer. After collecting data for a few seconds to obtain baseline fluorescence, purified 1 μM G α i-GDP was added to the solution to initiate GDP association. Once the fluorescence reached saturation, 10 \times (7.5 μM) GDP was added to the solution to initiate the GDP dissociation.

Molecular dynamics (MD) simulations

G α i-GDP coordinates were extracted from the G α i-G β γ complex structure (PDB: 1GP2)⁴⁰ with missing residues and atoms introduced using MODELLER 9v21.2⁴¹. The *pdb2gm*x module from the GROMACS-2020.3 package was used to generate both the 'charged' and 'uncharged' states for residues E236, D237 and E245⁴² to represent the 'charged' state as higher pH (above 7.2) and the 'uncharged' state as low pH (below 6.4). Following side-chain optimizations, minimum energy conformations of 'charged' and 'uncharged' G α i-GDP states were identified. Subsequently, MD simulations were performed to investigate conformational and dynamic differences compared to WT GDP-bound G α i. The CHARMM36 forcefield was used to parametrize the protein and GDP, and simulations were run using GROMACS-2020.3⁴³ (Supplementary Table 2). G α i proteins were solvated in a cubic box containing approximately 22,541 TIP3P water molecules and maintained a salt concentration of 150 mM through the addition of an appropriate number of Na^+ and Cl^- ions. The solvated systems were energy minimized using the conjugate gradient algorithm and

subsequently equilibrated for 1000 ns using the V-rescale thermostat and the Parrinello-Rahman barostat⁴⁴. Long-range electrostatic interactions were evaluated using the Particle-Mesh Ewald sum⁴⁵, and all bonds involving hydrogen atoms were constrained using the LINCS algorithm⁴⁶. Simulations were run for 1000 ns and were repeated in triplicate for reproducibility. Structural figures were generated using PyMOL³⁸ (<https://pymol.org/2/>) and UCSF ChimeraX³⁹ (<https://www.cgl.ucsf.edu/chimera/>). Graphical plots were created using Xmgrace (<http://plasma-gate.weizmann.ac.il/Grace/>).

MD simulations of the trimeric G α i-G β γ complex were initiated using the G α i-G β γ complex structure (PDB: 1GP2) as the starting point. The protonation states of residues E236, D237 and E245 were adjusted to simulate 'uncharged' ('0') and 'charged' ('-1') states, corresponding to different pH conditions. Additional MD simulations were also run by introducing the E236L + D237G + E245Q triple mutation into G α i to evaluate how changes in salt-bridge interactions within the Switch regions influence G α i-G β γ interactions. The MD simulations were run for 1000 ns in triplicate for robustness and reproducibility (Supplementary Table 2).

Molecular Mechanics Poisson-Boltzmann Surface Area (MMPBSA) analysis was performed using the *gm*x_MMPBSA tool⁴⁷ to compute the binding free energies between G α i and G β γ by analysis of contributions from electrostatic, van der Waals, and polar and non-polar solvation energies. In addition, residue-based energy decomposition from *gm*x MMPBSA was performed to evaluate pH-dependent contributions from individual residues to the overall binding energy of the trimeric complex.

Cell culture

HEK293T [American Type Culture Collection (ATCC), CRL-11268] cells were maintained, passaged, and transfected in Dulbecco's Modified Eagle Medium (DMEM, Gibco) containing 10% sterilized Fetal Bovine Serum (FBS, Gibco), 100 Units/mL penicillin, and 100 $\mu\text{g}/\text{mL}$ streptomycin (Gibco-ThermoFisher, Waltham, MA) in a humidified atmosphere at 37 $^\circ\text{C}$ and 5% CO_2 . Transfection was carried out using a lipid-polymer-based transfection agent (Mirus Bio, MIR 5400, Madison, WI). After transfection, cells were plated in DMEM (Dulbecco's Modified Eagle Medium) containing 10% FBS (Fetal Bovine Serum), 100 Units/mL penicillin, and 100 $\mu\text{g}/\text{mL}$ streptomycin for BRET assays.

Intracellular pH modulation

The intracellular pH of HEK293 cells was altered by two different methods. In the first method, HEK293 cells were treated with 1-5 μM of trifluoromethoxycarbonyl cyanide phenylhydrazide (FCCP) for 5 min. In the second method, intracellular pH was modulated by incubating HEK293 cells with Hanks' Balanced Salt Solution (HBSS) buffer at different pH for 15 min. Intracellular pH was quantified using the intracellular pH indicator, 2',7'-bis-(2-carboxyethyl)-5-(and-6)-carboxyfluorescein-acetoxymethyl ester (BCECF-AM) as described¹⁸. To convert BCECF-AM fluorescence to the intracellular pH values, HEK293 cells were treated with 20 μM nigericin and a calibration curve for pH_i and BCECF-AM fluorescence was generated.

Bioluminescence Resonance Energy Transfer (BRET) assays

BRET assays were carried out as described by Olsen et al.³⁴. Briefly, HEK293T cells were seeded in six-well dishes in DMEM containing 10% FBS, 100 Units/mL penicillin, and 100 $\mu\text{g}/\text{mL}$ streptomycin at a density of 300,000 cells/well and allowed to grow to 80-90% confluency. Cells were then transfected using a 1:1:1 DNA ratio of NTR:G α :RLuc:G β :G γ :GFP2 (1000 ng/construct for six-well dishes). TransIT-2020 (Mirus Biosciences, Madison, WI) was used to complex the DNA at a ratio of 3 μL TransIT/ μg DNA, in OptiMEM (Gibco-ThermoFisher, Waltham, MA) at a concentration of 1 μg DNA/ μL OptiMEM. The next day, cells were harvested from the plate using trypsin (0.25% EDTA, Gibco, #25200056) and plated in poly-D-lysine-coated white, clear bottom

96-well assay plates (Greiner Bio-One, Monroe, NC) in DMEM containing 1% dialyzed FBS, 100 Units/mL penicillin, and 100 µg/mL streptomycin at a density of 50,000 cells/well. One day after seeding in 96-well assay plates, white backings (Perkin Elmer, Waltham, MA) were applied to the plate bottoms, and the growth medium was aspirated. Cells were washed three times with Assay buffer (1X HBSS, 20 mM HEPES, pH 7.4), then 60 µL of Assay buffer was added immediately, followed by a 10 µL addition of freshly prepared 50 µM coelenterazine 400a (1-bisdeoxycoelenterazine) (Nanolight Technologies, Pinetop, AZ) in each well. After a five-minute equilibration period, cells were treated with 30 µL of 3x neurotensin (3×10^{-5} M) for another 5 minutes. Plates were then read in a plate reader (Clariostar, BMG Labtech, Ortenberg, Germany) at 410 nm (Rluc8-coelenterazine 400a) and 515 nm (GFP2) with a slit width of 8 nm using 10 flashes per spiral well scan. Plates were read serially five times, and measurements from the third read were used in all analyses. BRET ratios were computed as the ratio of the GFP2 emission to Rluc8 emission. For GFP control assays, Gβ and Gy-GFP2 (200 ng each) were transfected into HEK293 cells. Assay buffer (100 µL) was added following wash steps (lacking coelenterazine 500a or neurotensin) with fluorescence measured at 515 nm. The raw data was used to compare GFP fluorescence at different pH values.

Statistical analysis

CD melt-curves were fit to Boltzmann sigmoidal equation in Prism 9.3.1 (GraphPad Software, San Diego, CA). Concentration-response curves for BRET assays were fit to a three-parameter logistic equation. Raw BRET concentration-response curves were normalized to the best-fit maximum within a data set. BRET data were represented as mean ± SEM. Data analysis was carried out using Prism 9.3.1 (GraphPad Software, San Diego, CA).

Reporting summary

Further information on research design is available in the Nature Portfolio Reporting Summary linked to this article.

Data availability

PDB codes of previously published structures used in this study are: **1GP2** (GDP bound Gαi-Gβ-Gγ2), **1CIP** (GppNHp bound Gαi), **7SOF** (Isoproterenol bound beta1 adrenergic receptor in complex with heterotrimeric Gαi protein), **3OHM** (Gαq bound to phospholipase C beta 3), **1AZT** (GTPγS bound Gαs), and **1ZCA** (Gα12 in complex with GDP). Reference assignments were taken from **BMRB 30078** (Gαi subunit in complex with GDP). The co-ordinates of Gαi-GDP in 'charged' and 'uncharged' states in complex with Gβγ are provided as supplementary data files. All remaining data are available in the Article, Supplementary and Source Data files. All data reported in this paper will be shared by the lead contacts upon request. Source data are provided with this paper.

Code availability

Molecular dynamics simulations were performed using GROMACS-2020.3 (<https://www.gromacs.org>) with the CHARMM36 force field for protein and GDP parametrization. Simulations were carried out using standard GROMACS tools and the pdb2gmX module to prepare the 'charged' and 'uncharged' states of the Gαi protein. Structural and energy analyses were performed using the gmX_MMPBSA tool (<https://manual.gromacs.org>) for binding free energy calculations. Structural figures were rendered using PyMOL³⁸ (<https://pymol.org>) and UCSF ChimeraX³⁹ (<https://www.cgl.ucsf.edu/chimera/>). Xmgrace (<http://plasma-gate.weizmann.ac.il/Grace/>) was used for graphical plotting. The raw simulation data, scripts, and detailed protocols are available upon request. The MD simulation parameters files are available as Supplementary Data.

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Competing interests

The authors declare no competing interests.

Additional information

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