Use of High Pressure NMR Spectroscopy to Rapidly Identify Proteins with Internal Ligand-Binding Voids

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Abstract

Small molecule binding within internal cavities provides a way to control protein function

and structure, as exhibited in numerous natural and artificial settings. Unfortunately,

most ways to identify suitable cavities require high-resolution structures *a priori* and may

miss potential sites. Here we address this limitation via high-pressure solution NMR

spectroscopy, taking advantage of the distinctive nonlinear pressure-induced chemical

shift changes observed in proteins containing internal cavities and voids. We developed

a method to rapidly characterize such nonlinearity among backbone ^{1}H and ^{15}N amide

signals without needing to have sequence-specific chemical shift assignments, taking

advantage of routinely available ^{15}N-labeled samples, instrumentation, and 2D ^{1}H/^{15}N

HSQC experiments. From such data, we find a strong correlation in the site-to-site

variability in such nonlinearity with the total void volume within proteins, providing

insights useful for prioritizing domains for ligand binding and indicating mode-of-action

among such protein/ligand systems. We suggest that this experimental approach is a

rapid and useful probe of otherwise hidden dynamic architectures of proteins, providing

novel insights and opportunities into ligand binding and control.

Significance Statement

Many proteins can be regulated by internally binding small molecule ligands, but

it is often not clear *a priori* which proteins are controllable in such a way. Here we

describe a rapid method to address this challenge, using solution NMR spectroscopy to

monitor the response of proteins to the application of high pressure. While the locations

of NMR signals from many proteins respond to high pressure with linear chemical shift

changes, proteins containing internal cavities that can bind small molecule ligands

respond with easily identified non-linear changes. We demonstrate this approach on

several proteins and protein/ligand complexes, suggesting that it has general utility.

Introduction

Small molecule cofactors and ligands play critical roles in controlling protein

structure and function, the understanding of which often gives insight both into natural

and artificial modes of regulation. Of particular interest are identifying sites where the

irregular structures of proteins give rise to cavities, voids and other features which can

serve as internal sites for such compounds to bind (**Fig. 1**) and allosterically control

protein structure and function (1, 6, 11-13).

The traditional method to identify such locations – high-resolution X-ray

crystallography, ideally with better than 2 Å resolution to aid the particularly when

combined with computational analyses for cavity identification (5, 14-21) identification of

internally bound waters – can be powerful, or experimentally solving multiple structures

of proteins soaked with different organic solvents or small molecule fragments (22, 23).

Accordingly, this approach relies on having well-diffracting crystals, which are not

always available for all systems and can be time-consuming to produce even when

successful. In parallel, several classes of computational tools for predicting protein

cavities and other ligand-binding sites have rapidly advanced through a variety of

approaches (9, 21, 24-26). While these usefully expand the hypothesis space, they

also highlight the need for fast experimental validation.

As an alternative method to experimentally determine which proteins might

contain internal cavities suitable for ligand binding, we explored the potential for using

high pressure solution NMR to do so. We thought this approach might be useful, given

the need for proteins to undergo structural changes to allow for ligand binding within

their pre-existing internal cavities and voids (**Fig. 1**). While such changes may occur

Fig. 1: Example of internal ligand binding cavity and computational analysis. A.
Example of protein/ligand complex utilizing an internal cavity (HIF-2α PAS-B, PDB: 3F1O (6))
that is sequestered over 6 Å from solvent. In the apo- form of this protein (not shown), a pre-
formed 290 Å^3 cavity with 8 crystallographically-ordered water molecules is present at this site.
B. Schematic definitions of cavities, voids, and algorithm used by *ProteinVolume* (5) as used
for Fig. 3 and onwards in the manuscript. We refer to *cavities* as internal openings larger than
a single water molecule (V > 30 Å^3 (27)), while *voids* are more generally non-protein filled
spaces that include cavities along with other types of packing defects distributed throughout
and around a protein. With a protein structure available (gray), the <u>total void volume</u> can be
straightforwardly calculated by the difference of the solvent-excluded volume of a protein
(generated by rolling probes over the molecular surface, typically with water-sized radii) and the
volume taken by protein atoms = volume under blue surface minus gray volume (5).

rarely at ambient pressure, elevated pressure is known to dramatically increase the

equilibrium populations of the less populated, low-volume conformers associated with

hydration of cavities and voids (28). Such low-lying excited state conformers (N'), if

present, usually equilibrate with the ground state folded conformer (N) rapidly on the

NMR time scale (τ < ms). This gives rise to population-weighted single peaks in

multidimensional spectra which exhibit *non-linear chemical shift changes* as pressure

pushes the equilibrium N $\rightleftharpoons$ N' toward N' (29). We anticipated that this effect might be

easily detected in other globular proteins at pressures in the 500-2000 bar range, below

the pressures which typically partially or completely unfold proteins (29-31).

To evaluate this approach, we examined the effects of high hydrostatic pressure

on the NMR chemical shifts of a collection of protein domains and protein-ligand

complexes. Many of these proteins are members of the Per-ARNT-Sim (PAS) family of

ligand-controlled protein/protein interaction domains, which often internally bind different

small molecule cofactors to sensitize them to environmental factors like O_2, light, and

xenobiotics (32, 33). Changes to the occupancies or configurations of these cofactors

trigger conformational changes in the surrounding protein, regulating the activity of

various effector domains in natural and engineered proteins (33, 34). While a number

of high-resolution structures of apo- and ligand-bound PAS domains have been solved,

these models and predictions which can be made from them can provide insights into

only a small fraction of the many thousands of "orphan" PAS domains without known

ligands. Complementing these proteins, we added additional proteins and

protein/ligand complexes from a wide range of domain types – including some which

are known to bind ligands internally, some not – to establish the generality of this

approach for quickly probing protein structure and function.

Here we test the ability of high-pressure NMR to rapidly identify void-containing

proteins, with three key advances. First, by analyzing pressure titration data from over

40 proteins and protein/ligand complexes, we show that an easily accessible metric –

how differently sites within a protein respond to increasing pressure, as assessed by the

diversity of non-linear chemical shift perturbations observed in a simple titration without

requiring site-specific assignments – correlates well with the void volume within a

protein. We find that this metric is robust enough to predict total void volume on its own,

allowing the prioritization of potential ligand-binding capability amongst several targets.

Second, we demonstrate that internal ligand binding within a protein can reduce this

heterogeneity, quickly providing information on ligand binding and, in certain cases,

mode of action. Finally, we illustrate how this method can also be used to rapidly

assess the impact of point mutations/repacking, such as those used to fill or generate

cavities to facilitate artificial control, on the prevention or enabling of water entry. Taken

together, our data show the general utility of these rapidly acquired, easily analyzed

data to provide this important biophysical characterization of new proteins.

Results

Evaluation of non-linear NMR chemical shift responses to high pressure: We began by

examining the pressure dependence of NMR signals from a set of well-characterized

proteins using the workflow in **Fig. 2**. For each U-^{15}N labeled protein sample, we

acquired ^{1}H/^{15}N HSQC spectra at increasingly higher pressures from 20-2500 bar. After

each individual dataset was acquired at high pressure, we lowered the pressure to 20

bar and acquired a spectrum to confirm the reversibility of conformational changes; after

any sign of irreversible changes in peak intensity or location, the pressure series was

stopped. These series were typically composed of 21 spectra each taking 60 min

apiece, for a total of approximately 21 hr. Post-processing, peaks were picked and the

pressure dependence of changes of their ^{1}H and ^{15}N chemical shifts were

independently fit to a second-order polynomial equation:

Fig. 2: Workflow of pressure NMR analyses. For each protein analyzed, $^1H/^{15}N$ HSQC spectra were acquired at increasing pressures from 20-2500 bar, interleaving additional spectra at 20 bar between those at higher pressure to assess protein reversibility. Following processing, peaks were picked and chemical shifts monitored as a function of increasing pressure. Independently treating ^1H and ^{15}N movements, the pressure-dependent chemical shift changes of each peak are fit to the second order polynomial indicated to obtain linear (b_i) and nonlinear (c_i) coefficients. Subsequent analyses of the nonlinear (c_i) coefficients include either plotting c_i values as a function of residue number (for proteins with backbone chemical shift assignments) to identify regions with likely pressure-dependent conformational changes or to simply generate histograms of c_i values to give a quick initial characterization of likely ability to adopt multiple folded conformations.

$$\delta_i = a_i + b_i p + c_i p^2 \textbf{ (Eq. 1)}$$

As established by Akasaka and co-workers (29, 30), the linear (b_i) and nonlinear

(c_i) coefficients of these pressure responses reflect different properties of each protein.

To evaluate various analyses of these data, we examined an initial group of nine

proteins ("test set") with high-resolution structures with total void volumes ranging from

approximately 1500-8500 $\mathring{A}^3$ as assessed by ProteinVolume (5). This approach

provides a sum of the volumes of a wide array of packing defects, cavities, etc.

regardless of their size or distribution within a protein structure by simply calculating the

difference between the solvent-accessible volume of a protein and the volume occupied

by protein atoms (**Fig. 1B**). Pressure titration data were additionally recorded from

more than 30 additional proteins and protein-ligand complexes, with varying degrees of

structural information to build the "complete set" of data for subsequent analyses (a

complete list of all analyzed proteins and complexes is provided in **Table S1**).

As an initial analysis, we examined the absolute values of the two chemical shift

coefficients ($|b_i|$, $|c_i|$), separately averaged over all backbone amide protons and

nitrogens (typically 25-125) within each protein. From these analyses, we confirmed

that the averaged ^{1}H and ^{15}N $|b_i|$ values were fairly uniform across proteins with less

than two-fold variation, while the corresponding $|c_i|$ values varied over 4-10 fold ranges

(**Fig. S1**). These data expand the original data sets of Akasaka and Li (30) from seven

proteins and complexes to over 40 total, add an independent measurement and

analysis of the GB1 protein to assess the impact of different equipment, labs, and

software for data acquisition and analyses (GB1: 750 MHz in Kobe University by K.A.;

GB1' 700-800 MHz in New York by K.H.G.), and support the original interpretations that

the linear $|b_i|$ component is relatively fixed among systems while $|c_i|$ depends on a

protein-specific feature (30).

Further investigating these pressure-dependent chemical shift changes, we

noted a trend towards proteins with larger total void volumes having both higher c_i

values (**Fig. S1**) and a greater range of individual residue-specific c_i parameters than

proteins with smaller voids, both of which we thought might reflect a more heterogenous

structural response. While the larger c_i values are captured by the previously-described

average $|c_i|$ parameter (30), we examined several ways to quantitate the heterogeneity

in c_i values, settling on a combination of histograms and the standard deviation of the c_i

(stdev(c_i)). While the correlation between average $|c_i|$ and stdev(c_i) values for amide ^1H

and ^{15}N shifts (**Fig. S2**) suggest similarities between the two metrics, we opted to

proceed using stdev(c_i) to take advantage of the larger range of values provided by

dropping the absolute value operation. We also observed a high correlation between

the stdev(c_i) values of ^1H and ^{15}N (**Fig. S3**), suggesting that non-linear shift changes of

either nucleus report on pressure-induced changes despite differences in the structural

factors which influence them (35). Our subsequent analyses utilized stdev(c_i) of ^{15}N

chemical shift changes (= stdev($c_i[^{15}N]$)), which are thought to be most strongly

influenced by changes in backbone torsion angles (30, 35).

To examine the linkage of stdev(c_i) to total void volume, we measured the

nonlinear components ($c_i[^{15}N]$) of a test set of nine proteins (with one, GB1, duplicated

as noted above) with known structures and a range of total void volumes (**Fig. 3**).

Histograms of stdev(c_i) parameters exhibited the previously mentioned variability, with

Fig. 3: **Increasing diversity of non-linear pressure dependent chemical shift changes correlates with increased void volume**. Total void volumes calculated by ProteinVolume (5) for each protein are presented on the x-axis; the y-axis plots the standard deviation of the non-linear c_i parameters for the amide ^{15}N chemical shifts, using data from 25-125 crosspeaks for each protein. Note that the GB1 protein was analyzed from independently produced and acquired data from two labs, as noted in the text. The linear regression (red line, y= −1.564+0.0194x, omitting the lysozyme point) shows a correlation coefficient (r^2) of 0.929. The red dashed line indicates an arbitrary value of approximately 50 (x2x10^{-10}) ppm/bar^2 as a recommended minimum value for potential small molecule binding cavities, as noted in the text. Inset: histograms of measured ^{15}N c_i parameters for three selected proteins (using the same color scheme as main figure; histograms for all ten proteins provided in **Fig. S8**).

proteins having larger total void volumes typically having broader distributions than

those with smaller total void volumes (**Figs. 3 inset** and **S8**). We interpreted this trend

as stemming from voids enabling proteins to increasingly shift from the native N

conformation to a second folded N' conformation under pressure, reflected in the non-

linear chemical shift responses as the N' is progressively populated. Quantitating the

breadth of these distributions by stdev(c_i[^{15}N]) and plotting these versus total void

volume showed a linear correlation coefficient (r^2) of 0.929, supporting the linkage

between pressure-induced chemical shift nonlinearities to identify cavities. Of note, the

proteins with a stdev(c_i[^{15}N]) above 50 (x 2×10^{-10} ppm/bar^2) are all known to bind other

proteins or small molecules, suggesting an arbitrary value which could be useful to

prioritize for ligand-screening.

We note that the primary outlier of the observed linear correlation, hen egg white

lysozyme, is thermostable with a T_m of almost 75°C (36), likely hampering its transition

into an excited state under pressure unless its intrinsic stability of its basic folded state

is substantially lowered (*e.g.* by cooling close to the cold denaturation temperature (37)).

In addition, we cannot rule out potential contributions from lysozyme being an enzyme

instead of a signal transduction component, which may contribute to a different ability to

adopt alternative conformations reflected in pressure-induced non-linear chemical shift

changes. This remains to be explored more broadly in future studies.

Use of pressure NMR to reveal ligand binding mode of action – Well-defined cases:

Our correlation between increased heterogeneity of pressure-induced chemical shift

changes and void size makes a strong prediction that this route should provide a rapid

way to assess ligand binding: ligands which bind within the protein and reduce total void

volume should decrease the stdev(c_i) value of spectra recorded on the receptor. To test

this prediction, we evaluated how ligand binding affects the PAS-B domains of the

human HIF-2α and ARNT proteins, both of which are involved in the human hypoxia

response (38) and contain internal cavities known to bind artificial small molecule

ligands (1, 3, 6).

Prior studies of the HIF-2α PAS-B domain from us and others (1, 6, 39, 40) have

shown that it contains an internal water-filled 290 Å^3 cavity with no obvious access to

external solvent. However, a variety of screening efforts – from NMR-based fragment

binding screens to high throughput screens of HIF-2α/ARNT disruption (1, 2, 6, 40) or

protein stability (40) – have identified a wide range of small molecules which bind into

this cavity with nano- to micromolar affinities. High-resolution structures of these

complexes show that the ligands displace water and reduce HIF-2 activity by impacting

HIF-2α/ARNT interactions (41-43). To accommodate these binding events within a

solvent-inaccessible cavity, the HIF-2α PAS-B protein must dynamically fluctuate to

allow the entry of small molecules into the interior (44). Our pressure NMR analysis

confirmed this hypothesis, showing a very broad distribution of $c_i(^{15}N)$ responses with a

correspondingly large stdev($c_i[^{15}N]$) of 85 x (2×10^{-10} ppm/bar^2) (**Figs. 4A, S4**). We

anticipated that repeating these measurements in the presence of two nanomolar-

affinity compounds (2 (1) and 37 (2)) would show smaller non-linear chemical shift

changes than the apo protein given the smaller total void volume and expected reduced

flexibility of the protein/ligand complexes. Our data supported this, as we observed

decreases in stdev($c_i[^{15}N]$) from 85 x (2×10^{-10} ppm/bar^2) for the apo protein to 64 (2-

bound) and 53 (37-bound) x (2×10^{-10} ppm/bar^2), respectively. Notably, these decreases

do not solely correlate with loss of void volume in some cases: measurements

conducted on HIF-2α D1, a computationally-repacked variant with five point mutations

which reduce the cavity volume to 77 Å^3 (45) while retaining function, shows an

Fig. 4: High pressure NMR can provide mode-of-action information on ligand binding, even without structural information. (Left panel) **A**. Ribbon diagrams of HIF-2α PAS-B in its apo (PDB: 3F1P) and holo forms (complexed with 2, PDB: 4GHI), highlighting the location of the internal ligand binding cavity. Structures of two high-affinity inhibitors, 2 and 37 (1, 2) are also shown. **B**. Ribbon diagram of ARNT PAS-B (apo PDB: 4EQ1; KG-548 complex PDB: 8G4A) with the location of internal cavities, along with the structures of two moderate affinity binders, KG-548 and KG-655 (3). No structural details were available for a compound-bound form at the outset of this work. (Central panel) Histograms of the ^{15}N c_i coefficients measured on apo- and holo forms of HIF-2α and ARNT PAS-B domains using the approach diagrammed in **Fig. 2**. (Right panel) Comparisons of the stdev(c_i[^{15}N]) values of HIF-2α and ARNT PAS-B apo and ligand-bound samples, showing substantial effects upon binding small molecules (or point mutations, in the case of HIF-2α PAS-B "D1" variant) into the internal cavities.

increase in stdev(c_i[^{15}N]) up to 131 x (2x10^{-10} ppm/bar^2) (**Fig. 4A**). We speculate that

some dynamic or thermodynamic stability aspect of this engineered protein differs

markedly natural proteins, contribute to its unusual pressure response.

We used the same approach with the ARNT PAS-B domain, which has the same

fold as HIF-2α PAS-B but contains smaller internal cavities totaling 150 $Å^3$ in volume.

We have previously used NMR-based fragment screening, isothermal titration

calorimetry (ITC) and microscale thermophoresis (MST) to identify several compounds

which bind ARNT PAS-B with micromolar affinities; two of these, KG-548 and KG-655,

are further known to disrupt ARNT PAS-B interactions with coactivator proteins (3).

However, the binding modes of these compounds remained unclear without NMR or X-

ray co-complex structures available at the outset of this work.

To gain more insight on how KG-548 and KG-655 bound ARNT PAS-B, we used

our pressure NMR analysis, confirming that ARNT PAS-B retains a flexibly-accessible

interior cavity with a stdev(c_i [^{15}N]) of 84 x (2×10^{-10} ppm/bar^2) (**Figs. 4B, S5**). Repeating

these in the presence of the KG-655 ligand, we observed a drop of a stdev(c_i [^{15}N]) to

51 x (2×10^{-10} ppm/bar^2), consistent with an interior binding location for this small ligand

analogous to the HIF-2α PAS-B examples above. However, similar studies with the

larger KG-548 ligand show a stdev(c_i [^{15}N]) comparable to the apo protein, with an

observed value of 96 x (2×10^{-10} ppm/bar^2) that suggests that this compound binds in a

different mode than KG-655, potentially outside the cavity. We subsequently verified

this external binding mode for KG-548 by determining the X-ray crystal structure of

ARNT PAS-B with this ligand soaked into the crystal (**Table S2**). The resulting 1.97 Å

model (**Fig. S9**) clearly demonstrated that KG-548 bound across the outside of the PAS

β-sheet, which we independently validated in solution (46).

Use of pressure NMR to reveal ligand binding – Poorly-defined cases: As a next

demonstration of this approach, we examined its utility for PAS domains outside the

hypoxia response and with less well understood regulation. We considered two such

targets, the first of which is the N-terminal PAS domain of human PAS kinase (hPASK

PAS-A), a PAS domain-regulated serine/threonine kinase conserved among eukaryotes

(4, 47). This domain has a canonical PAS structure, with a five-stranded β-sheet

flanked to one side by several α helices (32, 48). While only a small surface groove of

73 $\mathring{A}^3$ could be identified near the F/G loop in the representative member of the solution

structure ensemble, an NMR-based fragment screen of 750 compounds identified

several small molecules that bound with micromolar affinity to hPASK PAS-A (4).

Chemical shift perturbations suggest that these compounds bind within the domain

interior, analogous to the subsequently discovered HIF-2α and ARNT PAS-B binding

compounds, but the lack of a hPASK PAS-A/ligand complex structure leaves this as an

open issue. To address this, we used high-pressure NMR to examine the response of

hPASK PAS-A in its apo form and when saturated with two compounds with K_D=13-24

µM affinities (KG-535, KG-571 = compounds 1 and 2 in (4)). For the apo- protein, we

observed a stdev(c_i[^{15}N]) of 45 x (2x10^{-10} ppm/bar^2); from the correlation identified in

our test set (**Fig. 3**), this value predicts that hPASK PAS-A contains a total void volume

slightly smaller than the 3345 $\mathring{A}^3$ calculated from the representative member of the

solution structure ensemble (**Figs. 5A, S6**). Upon addition of the KG-535 and KG-571

ligands, we observed smaller stdev(c_i[^{15}N]) values for both ligand-bound forms (35 and

28 x (2x10^{-10} ppm/bar^2) for the KG-535 and KG-571-bound forms, respectively (**Fig.**

5A). These data suggest that both compounds bind within the void volume of hPASK

Fig. 5: High pressure NMR can provide ligand binding information even in the absence of experimental structure. (Left panel) **A.** Three-dimensional structure of hPASK PAS-A (PDB: 1LL8) and structures of two binding compounds, KG-535 and KG-571 (4). No high-resolution structures of protein/ligand complexes are available, protein/ligand complex was generated using Boltz-2 (7). **B.** No experimental structural information is available for the RE137 PAS domain (8) (either apo or bound forms), structure shown here was generated using AlphaFold3 (AF3) (9, 10). Structures of two binding compounds identified by ^{19}F ligand-detected NMR screening, 15 and 47 and the protein/ligand complex was generated using Boltz-2 (7). (Central panel) Histograms of the [15]N c_i coefficients measured on apo- and holo forms of hPASK and RE137 PAS domains using the approach diagrammed in **Fig. 2**. (Right panel) Comparisons of the stdev(c_i[^{15}N]) values of hPASK and RE137 apo- and ligand-bound samples.

PAS-A, with KG-571 having a greater effect on the domain adopting alternative folded

conformations. Additionally, Boltz-2 (7) predicted holo structures of hPASK PAS-A

shows both ligands binding within an internal cavity (**Figs. 5B, S10B**), consistent with

our pressure NMR findings.

We completed our analyses of natural proteins by examining a prokaryotic PAS

domain from a novel histidine kinase from *Rhizobium etli (8, 49)*, chosen because of a).

homology with the light-regulated PAS-histidine kinase EL346 (50, 51) and b). the

widespread usage of small molecule ligands to control the structure and function of

PAS-containing proteins (including many histidine kinases (33)). While no experimental

structures are available of this domain, which we designate RE137, we hypothesized

that it may contain a cavity that can accommodate a small molecule, given precedence

from other PAS domains (33) and the presence of several small cavities within an

AlphaFold3 (AF3) (9) model of apo RE137 (**Fig. S10B**). Pressure NMR analyses gave

experimental support for this hypothesis, as the stdev(c_i[^{15}N]) value of 53 x (2×10^{-10}

ppm/bar^2) (**Figs. 5B, S7**) we obtained from the apo protein is consistent with a 2800 Å^3

total void volume from the correlation seen in **Fig. 3**. To provide an initial assessment

of small molecule binding to RE137, we used an R_2-filtered ^{19}F NMR assay to screen a

58-member library of fluorinated compounds (52) for their binding to this protein. Two

compounds from this screen showed binding, 15 and 47; subsequent pressure NMR

analyses of RE137 in presence of 1.2 mM of either compound exhibited decreases of

43% and 39% with stdev(c_i[^{15}N]) values of 30 and 32 x (2×10^{-10} ppm/bar^2) for the 15-

and 47-bound forms, respectively. We interpret these data to indicate that RE137 can

bind both compounds in a manner that reduces the total void volume of the protein,

consistent with interior binding locations for these small ligands.

Use of pressure NMR to probe artificially-designed proteins: As a final demonstration of

the utility of this method, we examined its utility in investigating the flexibility of an

artificially-designed protein, CA01 (53). With advances in protein engineering enabling

Fig. 6: High pressure NMR of a thermostable designed protein with an internal cavity gives smaller non-linear pressure dependent chemical shift changes than anticipated from void volume. (Left panel) Three-dimensional structure of CA01 (PDB: 5E6G), highlighting the location of a 75 Å^3 internal cavity within a total void volume of 3500 Å^3. (Right panel) Histogram of the ^{15}N c_i coefficients measured on CA01 (aqua) and 22 apo proteins listed in **Fig. S1** (grey) using the approach diagrammed in **Fig. 2**. Bar on right indicates stdev(c_i[^{15}N]) measured for CA01 (aqua) and 22 apo proteins (grey), along with the value predicted for CA01 from the correlation in **Fig. 3** (blue).

the development of artificial ligand-binding biosensor proteins (54), we sought to probe

whether a protein such as CA01 – which contains a completely-enclosed 75 Å^3 cavity

within approximately 3500 Å^3 of total void volume – might show similar pressure-

dependent non-linear chemical shift changes as natively-evolved counterparts. From

our correlation in **Fig. 3** we expected to see a stdev(c_i[^{15}N]) value of approximately 75

(2x10^{-10} ppm/bar^2) from a natural protein with void volume of this size, but we instead

observed a substantially smaller value of 38 x (2x10^{-10} ppm/bar^2) (**Fig. 6**). As CA01 is

extremely stable to thermal and chemical denaturation, requiring 5 M guanidinium

hydrochloride to exhibit a complete thermal melt with a 75°C T_m (53), we view this

observation supports our prior findings with lysozyme suggesting thermostable proteins

being less able to adopt alternative conformations (**Fig. 3**).

Discussion

In the present work, we successfully used non-linear pressure-induced chemical

shift effects at backbone amide ^{1}H and ^{15}N nuclei to rapidly assess whether proteins

contain substantial total void volumes within them and if these can bind small molecule

ligands. A particular strength of our implementation is its simplicity, as 1). U-^{15}N labeled

samples are routinely available from *E. coli* or *in vitro* expression systems (or otherwise,

using natural abundance ^{15}N samples with optimized data collection), 2). NMR data

needed are easily obtained from conventional ^{1}H/^{15}N HSQC experiments recorded at

variable pressures using commercially available pressure equipment, and 3).

straightforward data analyses are done by simply tracking peak locations without

requiring chemical shift assignments, eliminating a time-consuming step of NMR

analysis. These characteristics make our method broadly accessible, data-rich, and

fast enough to integrate with emerging discovery pipelines.

Mechanistically, we attribute the non-linear chemical shift changes with

increasing pressure to site-specific conformational changes of the protein from the

native folded structural ensemble (N) to the low-lying excited state ensemble (N') with a

smaller volume (17). Our specific choice of backbone amide ^{1}H and ^{15}N chemical shifts

give us complementary views of these transitions (**Fig. S3**), as ^{1}H chemical shifts are

dominated by hydrogen bonding to these atoms, while ^{15}N chemical shifts are

influenced by a broader set of factors, including changes in hydrogen bonding (to both

the adjacent ^{1}H and carbonyl CO groups), local backbone dihedrals and sidechain

conformations (55). Of note, non-linear pressure-dependent chemical shift changes at

these sites are much smaller in peptides or intrinsically disordered proteins (56).

Indeed, non-linear pressure-dependent amide ^{1}H and ^{15}N chemical shift changes have

long been recognized as reflecting idiosyncratic differences among proteins (57). Prior

work by one of our groups (K.A.) suggested that this property depends on the density of

cavities within the protein (30); here we both expand the number of proteins examined

from that work and suggest that contributions likely arise from small and large voids

distributed throughout the protein which collectively give rise to the total void volumes

calculated by the program ProteinVolume (5) given the excellent correlation we present

in **Fig. 3**. We posit that such an approach reveals aspects that will be correlated to the

ability of proteins to be similarly affected by allosteric regulators and other ligands,

linking these non-linear chemical shift effects to the potential for small molecule binding

and regulation.

In addition to getting information on the total void volume within a given protein,

the strengths of this analysis include the ability to identify ligands which bind to a given

target and (de)stabilize one conformation in a way that may provide allosteric switching.

As such, this provides trivially accessible mode-of-action information that is often

challenging to get otherwise without resources such as compounds of known binding

location for competition studies or cavity-filling point mutants. Our comparison of two

cavity-perturbing effects for HIF-2α PAS-B – small molecule binding and cavity-filling

redesign – produced opposite effects, with the small molecules stabilizing the native fold

and mutations destabilizing in our pressure NMR analyses. We underscore that these

effects occur despite minor changes in ground state structure among structures of the

apo, ligand-bound and redesigned proteins (1, 6, 45). As such, pressure-dependent

chemical shift effects give us insight into details of the packing and dynamics of proteins

more simply than most other methods can provide. Finally, our comparison of the

ARNT PAS-B binding compounds KG-548 and KG-655 provided an outstanding

example of the power of the pressure NMR method to quickly and accurately provide

mode-of-action binding site information even with limited structural data available. In

these cases, high pressure NMR – using rapidly-acquired spectra without needing any

chemical shift assignments – indicated different binding modes between the two

ligands, with KG-655 binding primarily to an internal site while KG-548 bound externally.

These findings were subsequently validated both by a ARNT PAS-B:KG-548 crystal

structure (**Fig. S9**) and accompanying solution biophysical investigation (46).

We emphasize that our approach requires very little preliminary information on

the target, needing neither structures nor site-specific chemical shift assignments, as

demonstrated by the RE137 case. In this case, the stdev(c_i[^{15}N]) values of the apo

protein exceeds the value of approximately 50 x (2×10^{-10} ppm/bar^2) value that we

suggest as indicating the potential for ligand binding, and we were able to quickly

identify compounds which bound RE137. Our high pressure NMR results quickly add

important information, showing that these compounds bind and modulate various

aspects of protein structure and dynamics, probably through binding internal cavities.

A final strength of the method is its speed, making this pressure NMR method an

outstanding complementary experimental tool to computational pipelines which can

predict protein cavities (21) or protein ligand interactions (9, 24-26) at scale. While

powerful, these predictions require experimental validation to confirm which proposed

interactions translate into meaningful binding events in solution. Via ligand-induced

changes in stdev(c_i[^{15}N]) values, pressure NMR provides a rapid experimental screen to

identify proteins and small molecule binders worth pursuing with higher resolution

experimental NMR, X-ray, or cryo-EM analyses.

Finally, while our demonstrations have focused on relatively small ~15 kDa ligand

binding domains, we note that several routes can expand the reach of this approach to

larger and more complex systems by reducing spectral complexity. Options include

biochemical approaches to generate proteins with simplified ^{15}N-labeling patterns, as

provided by amino-acid type selective or sortase/intein-based domain-specific

techniques, to ^{1}H/^{15}N TROSY or HNCO-type pulse sequences which can provide

improved resolution. Additional options may be provided by the improved NMR signal

properties of ^{1}H/^{13}C methyl signals for similar pressure-sensitive studies, particularly

given the placement of these residues near internal cavities.

Materials and Methods

Proteins were expressed in *E. coli* with uniform ^{15}N labeling, purified using a

combination of Ni(II) affinity and gel filtration chromatography before being exchanged

into a barostatic Tris:phosphate buffer mix which limits pH changes during

pressurization (58) and concentrated to 100-700 µM for NMR spectroscopy. ^{1}H/^{15}N

HSQC spectra were acquired at increasing pressures from 20-2500 bar, interleaving

each high pressure spectrum with a low pressure (20 bar) spectrum to confirm that

changes in peak locations and intensities were reversible. All NMR data were

processed and analyzed with NMRviewJ (59, 60) and NMRFx (59, 61, 62). After

individually processing each spectrum, we picked peaks and tracked their changes in

chemical shifts as a function of pressure. Separately handling movement in the ^{1}H and

^{15}N dimensions, we fit these trends to a second-order polynomial equation (**Eq. 1, Fig.**

2). Protein volume analyses of single isolated cavities utilized *cavfinder* (1), while

measurements of total void volumes utilized *ProteinVolume* (5). Additional detailed

procedures are found in **SI Materials and Methods**.

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A. HIF-2α PAS-B

internal cavity

B. voids

probe

surface groove

A.

apo
(3F1P)

2

37

B.

apo
(4EQ1)

KG-548

KG-655

KG-548

HIF-2α PAS-B (apo)

HIF-2α PAS-B (2)

HIF-2α PAS-B (37)

population %

c_i (^{15}N) (2x 10^{-10} ppm/bar^2)

ARNT PAS-B (apo)

ARNT PAS-B (KG-548)

ARNT PAS-B (KG-655)

population %

c_i (^{15}N) (2x 10^{-10} ppm/bar^2)

stdev of c_i (^{15}N)

HIF-2α

D1

apo

stdev of c_i (^{15}N)

ARNT

KG-548
apo

KG-655

A.

apo
(1LL8)

KG-535

KG-571

KG-571

hPASK (apo)

hPASK (KG-535)

hPASK (KG-571)

c_i (^{15}N) (2×10^{-10} ppm/bar^2)

population %

PASK

stdev of c_i (^{15}N)

apo
KG-535
KG-571

B.

apo
(AF3)

15

47

RE137 (apo)

RE137 (15)

RE137 (47)

c_i (^{15}N) (2×10^{-10} ppm/bar^2)

population %

RE137

stdev of c_i (^{15}N)

apo
15
47