Dimerization and dynamics of angiotensin-I converting enzyme revealed by cryo-EM and MD simulations

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<u>Abstract</u>

Angiotensin-I converting enzyme (ACE) regulates the levels of disparate bioactive
peptides, notably converting angiotensin-I to angiotensin-II and degrading amyloid beta.
ACE is a heavily glycosylated dimer, containing 4 analogous catalytic sites, and exists in
membrane bound and soluble (sACE) forms. ACE inhibition is a frontline, FDA-approved,
therapy for cardiovascular diseases yet is associated with significant side effects,
including higher rates of lung cancer. To date, structural studies have been confined to
individual domains or partially denatured cryo-EM structures. Here we report the cryo-EM
structure of the glycosylated full sACE dimer. We resolved four structural states at 2.99
to 3.65 Å resolution which are primarily differentiated by varying degrees of solvent
accessibility to the active sites and reveal the full dimerization interface. We also
employed all-atom molecular dynamics (MD) simulations and heterogeneity analysis in
cryoSPARC, cryoDRGN, and RECOVAR to elucidate the conformational dynamics of
sACE and identify key regions mediating conformational change. We identify differences
in the mechanisms governing the conformational dynamics of individual domains that
have implications for the design of domain-specific sACE modulators.

Introduction

Angiotensin-I converting enzyme (ACE) is an M2 clan zinc metalloprotease (EC3.4.15.1) located on the plasma membrane and in the extracellular milieu (1-3). ACE plays a crucial role in the renin-angiotensin system by converting angiotensin I to the potent vasoconstrictor angiotensin II, and in the kinin-kallikrein system by degrading bradykinin, a potent vasodilator. Inhibiting ACE is a primary approach for treating hypertension and is a proven therapy for heart failure, diabetic nephropathy, and other cardiovascular and renal dysfunctions (4). However, ACE inhibition has known side effects, e.g., dry cough and angioedema, and has been linked to higher rates of lung cancer (5, 6). This is partly because ACE also degrades a diverse range of peptides, resulting in complex physiological outcomes due to the interplay in altering levels of these peptides (4, 7-10). Furthermore, ACE degrades amyloid β (Aβ), which is associated with the progression of Alzheimer's diseases (11-14). Overexpression of ACE in myelomonocytes substantially reduced Aβ load and prevented Alzheimer's disease-like cognitive decline in mice (15). However, the brains of some Alzheimer's disease patients showed upregulation of the renin-angiotensin system which led to deleterious effects and ACE inhibition was beneficial (16-18). These studies indicate that global ACE inhibition could be either beneficial or harmful for individual Alzheimer's patients, and underscores a need for better understanding the molecular mechanisms governing ACE function and substrate selectivity to develop more selective, improved, ACE inhibitors and expand ACE-based therapies.

ACE is a 180 kDa heavily N-linked glycosylated type 1 transmembrane protein with two homologous catalytic domains residing at the N- and C-terminal ends (ACE-N and ACE-C) in its extracellular region (1, 3, 19). The ACE extracellular region can be released via proteolytic cleavage near the transmembrane region by ADAM (a disintegrin and metalloproteinase) family protease(s), generating soluble ACE (sACE) and ACE dimerization can regulate this process (20-22). ACE belongs to the cowrin (cowry like) family of zinc metallocarboxypeptidases because its catalytic domain is ellipsoidal and contains a long, deep, yet narrow catalytic cleft (23). ACE is known to cleave short peptides, such as angiotensin I (decapeptide) and bradykinin (nonapeptide) by removing the C-terminal two amino acid residues. Different from the funnelin family of metallocarboxypeptides that contains a funnel-like cavity, the catalytic cleft of ACE can

accommodate entire oligopeptides and ACE has been shown to cleave multiple discrete

sites on Aβ outside the C-terminal end (13). Thus, it is more accurate to refer to ACE as

an endopeptidase with dipeptidyl carboxylase activity.

Extensive crystallographic studies of ACE-N or ACE-C alone and in complex with

various inhibitors and a catalytic product, angiotensin-II, have elucidated the overall

structures of ACE catalytic domains and molecular details of the catalytic pockets (2, 24-

28). The structures indicate that each domain is comprised of three sub-domains and

offer insight for the substrate preferences around cleavage sites and the promiscuity of

each domain toward short peptide substrates. Furthermore, they reveal the orientation of

substrate binding sites around scissile bond, indicating that the entire oligopeptide

substrates need to be engulfed into the catalytic chamber of ACE catalytic domain (24,

26, 27).

However, the catalytic pocket of ACE in the closed state is not large enough to

accommodate the entrance of all its known substrates, particularly Aβ, to permit the

observed cleavage pattern. Thus, similar to other Aβ-degrading proteases, such as insulin

degrading enzyme (IDE), presequence protease (PreP), and neprilysin (NEP), ACE must

undergo a large-scale open-closed transition to capture and degrade larger peptide

substrates such as Aβ (29-31). Recent open state structures of sACE-N, sACE monomer,

and a sACE-N dimer, along with molecular dynamics (MD) simulations of sACE-C, have

begun to reveal ACE conformational dynamics, though they remain under-studied (32-

34).

The extracellular domains of ACE are readily dimerized in the membrane-bound

and soluble forms. Both the ACE-N and ACE-C domains within the ACE monomer are

catalytically active and each has its own substrate preferences (7, 13, 35). For example,

ACE-N can convert Aβ1-42 to Aβ1-40, which is less amyloid fibril-prone than Aβ1-42 but

ACE-C cannot (13). Accumulating evidence support a complicated interplay between

ACE-N and ACE-C within the ACE dimer. Enzyme kinetics analysis suggests negative

cooperativity between two catalytic domains (36-38). However, ACE also exhibits positive

synergy toward Aβ cleavage and allostery to enhance the activity of its binding partner,

bradykinin receptor (13, 39). Recently, cryo-EM structures of soluble dimeric apo-sACE

were reported (33); however, due to the poor resolution/denaturation of sACE-C, only full-

length sACE monomer or dimeric sACE-N was reported yet the conformations preclude

full-length sACE dimerization due to the severe steric crash between sACE-C domains.

Here, we present four discrete cryo-EM structures of full-length sACE that contain all four

catalytic domains (Fig. 1A). This is followed by cryo-EM heterogeneity analysis and all-

atom MD simulations. Together, our analyses provide structural insights into domain

interactions, open-closed transitions of the catalytic domain, and conformational

dynamics of soluble dimeric ACE.

Results

Structures of the full-length sACE dimer determined by single particle cryo-EM

Soluble human ACE (sACE), which includes the entire extracellular region of ACE,

was expressed in HEK293F cells and purified from culture supernatant using an ion

exchanger (Source Q) followed by size exclusion chromatography (Superdex S200).

Buffer conditions suitable for single particle cryo-EM analysis were identified and

optimized using differential scanning fluorometry (Figure S1, Table S1). Purified sACE

was then vitrified using a Vitrobot, and a dataset of ~3,600 micrographs was collected on

a 300 keV Titan Krios equipped with a Gatan K3 camera at the University of Chicago

Advanced Electron Microscopy Facility (Table S2). The data was processed in

cryoSPARC v4.4.0 (Figure S2). After filtering ~1.1 million picked particles to ~378K using

2D and 3D classification, three classes were identified (Figure 1, Figure S2). Of these,

~268K particles mapped to a dominant 2-domain class reconstructed as an sACE-N

dimer, while ~21K particles mapped to a minor 3-domain class containing two sACE-N

and only one sACE-C, reminiscent of the recently published sACE cryo-EM structure (33).

The remaining 89K particles mapped to a full-length sACE dimer containing two sACE-N

and two sACE-C, which was refined to 3.65Å resolution (Figure 1C, S2, S3, Movie S1,

Table S2).

The partial denaturation of macromolecules during vitrification has been attributed

mainly to repeated exposure of the macromolecules to the air-water interface (40, 41).

The sACE-C domain was shown to be less stable than the sACE-N domain (42), so we

hypothesized that repeated exposure to the air-water interface could cause preferential

denaturation of sACE-C. We have previously showed that preferential denaturation of

human PreP occurred in one of its two homologous domains and that faster vitrification

using Chameleon significantly prevented such preferential denaturation (29). Therefore,
we collaborated with the National Center for CryoEM Access and Training (NCCAT) to
prepare sACE grids using Chameleon (43). Chameleon preparation reduced the time the
sample spends on the grid before freezing by more than 10-fold. We collected a new
dataset of ~18,000 micrographs and our 2D classification revealed a higher percentage
of particles for four-domain classes than those prepared using the Vitrobot (Figure 1B).
However, our 3D classification still showed the existence of structures of the sACE-N only
dimer and those containing two sACE-N and one sACE-C (Figure S4). This indicates that
the reduction in vitrification time using Chameleon reduced but did not prevent preferential
denaturation.

From these data, we obtained a map for a consensus ACE structure that achieved
a resolution of 2.8Å. However, while the core region showed excellent map quality, the
map quality for the dynamic regions corresponding to the open-close transition of ACE
catalytic domains was poor, suggesting multiple discrete conformational states exist. We
thus performed 3D classification and reconstructed three distinct conformational states of
full-length apo-ACE at resolutions of 2.99Å, 3.05Å, and 3.15Å (Figure 1C, S4, S5, Movie
S2-4, Table S2). We refer to these ACE dimer structures by their resolutions, namely
sACE-2.99, sACE-3.05, and sACE-3.15, along with the structure from data obtained using
the Vitrobot, sACE-3.65. Given the rich structural information from more than 50 reported
crystal structures of ACE and one recently reported cryo-EM study of ACE, below we
focus on the new insights that our sACE dimer structures offer (2, 24, 26, 33, 44, 45).

The overall structures of sACE dimers and structural insight into dimerization

Soluble ACE has two homologous catalytic domains: sACE-N (amino acids 1-615)
and sACE-C (amino acids 616-1202). Based on the comparison of various catalytic
domains of our ACE dimer structures, we subdivide each catalytic domain of sACE into
three subdomains: D1a, D1b, and D2 for sACE-N, and D3a, D3b, and D4 for sACE-C
(Figure 1A). These three subdomains in either sACE-N or sACE-C form a catalytic
chamber. The catalytic zinc ion is coordinated by the canonical HEXXH motif residing in
the D1b and D3b subdomains (Figure 1A). Four sACE dimer structures manifest as
pseudo-symmetric C2 homodimers across the dimer interface. The dominant symmetry-
breaking features are the degrees of displacement between D1 and D2 or between D3
and D4, leading to changes in solvent accessibility to the catalytic pockets.

Based on the openness of the catalytic chamber, defined as the distance between
the D1/3a and D2/4 helices that gate access, we classified sACE-N into three
conformational states: open (O), intermediate open (I), and closed (C) (Figure 2A); sACE-
C is significantly less open than sACE-N in our structures, and is only classified into
intermediate (I) and closed (C) states (Figure 2B). The four sACE dimer structures differ
mostly by the degrees of openness of their four catalytic domains (Figure 2C, D). Our
sACE-2.99 and sACE-3.65 structures are globally similar, both contain sACE-N domains
in the open and intermediate states and sACE-C domains in the closed and intermediate
states. In sACE-3.05, both sACE-N domains adopt an open state, while the sACE-C
domains adopt an intermediate and closed state. In the sACE-3.15 structure, one subunit
has both sACE-N and sACE-C adopting a closed conformation, while the other subunit
has an open sACE-N domain and an intermediate sACE-C domain.

Our four cryo-EM structures of sACE provide the structural basis for sACE
dimerization. sACE dimerization is mediated by interfaces between D2 subdomains and
D4 subdomains of sACE monomers. The D2 and D4 subdomains generally have much
lower B-factors, and higher local resolution estimates, than the D1 and D3 subdomains
(Figure S3, S5, S6). The interaction of the D4-D4 subdomains is analogous to the D2-D2
interaction, yielding an RMSD of ~1 Å² when aligned (Figure 3A). The primary source of
variation is a general rigid body shift away from the interaction interface for non-interacting
regions in the D4-D4 region compared to the D2-D2 region. The observed D2-D2 interface
is consistent with one of three possible contacts between sACE-N in the non-
crystallographic symmetry mates from previous crystallographic analysis of sACE-N and
the previously published cryo-EM structure (Figure S7). In total, the interface between
sACE monomers buries a surface area of approximately 1,900 Å². Although sACE-N and
sACE-C are structurally analogous, their respective interfaces (sACE-N/N and sACE-
C/C) are not equivalent. The sACE-N/N interface is larger and stronger, burying a surface
area of approximately 1,150 Å², compared to a buried surface area of approximately 750
Å² for the sACE-C/C interface.

The interface between sACE-N is formed by a hydrophobic core, anchored by a
Y465 π-stacking interaction, and flanked by salt bridges formed by R453 and R458 as
well as E212 and E219 along with a collection of polar and van der Waal contacts, which
is consistent with the previous mutational studies of Y465 (Figure 3B) (46). The dimer

interface between two sACE-N domains is completed by glycan-glycan interactions between the glycans attached to N82, which is consistent with the role of glycosylation in sACE dimerization (Figure 3C) (47). However, the central residues that form the hydrophobic core of the interface between sACE-N are not present in the interface between sACE-C. Instead, the sACE-C interface is formed by a symmetric salt bridge between K1067 and E1060, side chain-backbone interactions, and van der Waals contacts (Figure 3D).

Conformational states of sACE catalytic domains

The comparison of our dimeric sACE cryo-EM structures reveals the conformational dynamics of sACE catalytic domains. The four sACE dimer structures represent different combinations of open, intermediate, and closed states of sACE catalytic domains (Figure 2). By aligning each sACE-N and sACE-C region in our structures to the D2/4 domain, it becomes clear that our structures represent a continuous gradient of structural heterogeneity, ranging from the most closed state to the most open state (Figure S8).

We performed a series of all-atom molecular dynamics (MD) simulations to explore the conformational dynamics of sACE. A total of 8 apo-sACE simulations were conducted, comprising four simulations with non-glycosylated sACE (>1.8 μs each) and four with sACE where 12 N-linked glycosylation sites that contained observable extra density were modeled with complex glycans (>2 μs each) (Movie S5, S6). Glycan orientations were randomized prior to each simulation to avoid bias. We did not observe a substantial difference in the global conformational dynamics between our sets of simulations with and without glycans. The domain motions of these simulations are summarized in Table S3.

RMSD analysis reveals that the dominant source of conformational change in our simulations is the open-close transition between the D1/3 and D2/4 domains (Figure S9). The D2/D2 and D4/D4 regions are the most stable, with the bulk of conformational changes localized to the D1/D3 domain. Within the D1/3 domain, we observed that the D1/3a motion was not always correlated with D1/3b or D2/D4, meaning that at a given time, D1/3a could move along with D1/3b, or with D2/D4, or independently. Consistent with our structural analysis, our MD simulations also indicate that sACE-N displays a much wider range of opening geometries than sACE-C.

Most of our MD simulations rapidly closed, and remained closed for the duration
of the simulation, preventing an effective comparison of the open vs closed state
dynamics. However, we were able to garner significant insight into the mechanism of
open-close transition by integrating our MD simulations and cryo-EM structural analysis.
Initially, we sought to understand the stable and mobile regions of sACE by calculating
the Cα displacement for each residue among the four difference cryo-EM structures. As
shown in Figure 4A, the D2 domain was found to be the most stable region of sACE while
the D1 domain was the most mobile, highlighting the impact of the sACE-N open-close
transition. sACE-C was, globally, very stable outside of surface exposed loops and the tip
of the D3a subdomain. This analysis predicted the most mobile regions of sACE to be the
ends of the D1a subdomain and a loop region with the D1b that closely associates with
the tip of the D1a subdomain.

While the structural perturbations predicted by comparing our cryo-EM structures
correlate well with the global RMSD analysis of our MD simulations (Figure S9), we
performed further analysis to understand the finer details of sACE conformational
dynamics. To best compare the cryo-EM structures with our MD simulations, we
compared the calculated alpha carbon displacement of our cryo-EM structures to the root
mean square fluctuation (RMSF) values for each residue in our MD simulations (Figure
4). We observed consistent, global agreement between our cryo-EM data and MD
simulations, both with and without glycans present. This consistency is particularly
noticeable in several key mobile regions (denoted 1-4 in Figure 4) that demonstrate high
degrees of Cα displacement from our cryo-EM data along with correlating higher RMSF
during subsets of our MD simulations (Figure 4A, B, D). The RMSF variation between
stable and mobile regions in our MD simulations is dampened, relative to the cryo-EM
data when the entire trajectories were used (Figure 4A, C, E). This is due to the open
state regions closing relatively early in the simulation and remaining closed for the
duration of the simulation. We then performed principal component analysis (PCA)
(Figure S10) and cluster analysis (Figure S11) to compare our MD simulation data with
our cryo-EM data. Both methods of analysis indicate that the conformational space
sampled by our cryo-EM structures represents a fraction of the conformational space
sampled by our MD simulations.

The dynamics observed in our MD simulations of sACE are highly complex, arising

from a combination of both interdomain and intradomain motions. To summarize these

simulations in simpler terms, individual sACE domains transition from a more open state

to a closed state without reopening, while sACE-N/N and sACE-C/C primarily move

relative to each other as rigid bodies. As outlined previously, PCA was employed to reduce

the complexity of these motions and to identify dominant trends in conformational

changes within the dataset. Visualizing the conformational changes captured by the

principal component vectors in the MD simulations (Movies S7, S8) revealed that the

majority of conformational variability within the MD data is driven by substantial

interdomain motion occurring between sACE-N/N and sACE-C/C. Consistent interdomain

motion was observed along all principal component vectors, regardless of the presence

or absence of intradomain motion. However, no principal components were identified that

exclusively described significant intradomain motion without associated interdomain

motion. The PCA further revealed that glycosylation significantly enhances the degree of

interdomain motion in sACE, as evidenced by the glycosylated simulation (Movie S8)

compared to the non-glycosylated simulation (Movie S7). This suggests that glycosylated

sACE exhibits greater global dynamics (Figure 4, S12). Notably, principal components 5

and 6 from the glycosylated MD simulations (Movie S8) highlight a striking difference,

where sACE-N/N and sACE-C/C appear to move independently and exhibit a decoupled

interaction. We postulate that this inter-domain motion may be the reason we observe

preferential denaturation of sACE-C during cryo-EM grid preparation. The relatively harsh

conditions that the protein is exposed to during the grid-making process may damage

specific conformations of particles and subsequently restrict the amount of information

that could be obtained solely from cryo-EM analysis. This highlights the importance of

integrating multiple approaches to investigate protein dynamics.

The MD simulations of glycosylated sACE provides additional structural insights

into the structure and functions of N-linked glycosylation of sACE. Glycosylation is shown

to affect the folding, secretion, and stability of ACE (22, 48, 49). Our MD simulation reveals

that the dynamics of 12 N-linked glycans cover a large surface area of sACE, shielding it

from environmental insults (Figure S13). Consistent with observations from our Coulomb

potential density maps and previously published data, we observed a substantial and

persistent inter-subunit interaction between glycans from N82 (Figure 3C). From RMSF

analysis, glycosylation was found to increase the overall dynamicism of sACE (Figure
S12). Furthermore, while there are N-linked glycans near the catalytic chamber,
particularly those at the tip of D1a and D3a, they were highly dynamic and did not occlude
access to the catalytic chamber. This is consistent with the observation that
deglycosylation only modestly increases the enzymatic activity of ACE (50-52).

The comparison of sACE-N in its open and closed states reveals the structural
basis for the open-to-closed transition of this domain. Our structures show that when the
bottom of the D1a subdomain (region of interest 2 in figure 4) moves away from the D2
subdomain, resulting in the open state, the top of the D1a subdomain (region of interest
1 in figure 4) moves closer to the D2 subdomain, as if the helix were balanced upon a
fulcrum (Figure 5A, B). Detailed structural analysis indicates that the fulcrum is primarily
formed by the salt bridge between two highly conserved residues: E66 in the D1a
subdomain and R108 in the D2 subdomain. To stabilize the open state, additional salt
bridges are formed, with D189 in the D2 subdomain pairing with K73 in all cases and
occasionally with R96 in the D1a subdomain (Figure 5A). The loss of this salt bridge in
the closed state is compensated for by the formation of a network of van der Waals
interactions between the D1a and D2 subdomains that stabilize the closed state. The
closed state of sACE-C is also stabilized by a network of van der Waals contacts between
the D3a and D4 subdomains. The transition between open and closed states is likely
influenced by the binding of substrates and reaction products during the ACE catalytic
cycle. Although a similar fulcrum mechanism is not evident in the sACE-C domain, the
connection between the D3a and D2 subdomains would act as a lever to pivot sACE-C
from the closed to the open state, and we speculate that this may form the structural basis
for allosteric communication between sACE-N and sACE-C.

On average, sACE-N is more open than sACE-C, likely due to multiple factors
(Figure 2D). In sACE-N, the D1a subdomain is at the N-terminal end of sACE, allowing
for a greater range of motion (Figure 5B), while the fulcrum and stabilization mechanisms
can maintain sACE-N in a more widely open state (Figure 5A). Conversely, in sACE-C,
the motion of the D3a subdomain is constrained by its connection to the D2 subdomain
(Figure 5C). Additionally, there are sequence differences at the interface between the
three subdomains of sACE-N and sACE-C. In the most closed state—the B chain sACE-
C domain of the sACE-3.05 structure—residues from all three subdomains, including

I649, V724, V926, and V927, form a hydrophobic latch that stabilizes the closed state

(Figure 5D). The hydrophobic latch in sACE-N is weaker because the residue

corresponding to V724 has been replaced with a threonine (T124) (Figure 5D). These

structural insights are supported by our RMSF analysis of our MD simulations and alpha

carbon displacement analysis of our cryo-EM structures, which reveal a correlation

between mobile regions and the "fulcrum" mechanism of opening for sACE-N (Figure 4,

5A). Furthermore, our slowest closing MD simulation revealed that the K73-D189

interaction was maintained for the entire time that sACE-N adopted an open

conformation, yet this interaction was lost when sACE-N transitioned to the closed state

(Figure 5E).

We also performed three MD simulations on the non-glycosylated sACE monomer

(>1 microsecond each) (Table S3). We observed the expected intra-subdomain open-

closed motions in sACE-N and sACE-C that were also observed in the simulations of ACE

dimer. We also observed that the motion of sACE-C relative to sACE-N was significantly

greater in our monomer simulations compared to our dimer simulations. This suggests

that dimerization places significant constraints upon the conformational dynamics of

sACE. Previous work has defined the conformational dynamics of sACE from cryo-EM

data where both sACE-C domains were not resolved (30). Given our observations from

the monomeric and dimeric MD simulations, we endeavored to understand the

conformational heterogeneity present within our cryo-EM data to better understand the

impact of dimerization on sACE dynamics.

Cryo-EM heterogeneity analysis of sACE

Cryo-EM structures are ensemble averages of many aligned particle images.

Recently, there have been many approaches developed to understand the particle

heterogeneity present within a cryo-EM structure as a proxy for molecular motion (53).

We employed three methods of heterogeneity analysis: cryoDRGN, RECOVAR, and 3D

Variability Analysis (3DVA) (54-56). CryoDRGN employs neural networks to resolve

structural heterogeneity in a nonlinear, multidimensional latent space, while RECOVAR

and 3DVA both employ linear subspace methods. 3DVA finds the linear subspace by an

alternating optimization scheme to capture image variations while RECOVAR utilizes a

regularized covariance estimator to identify subspaces that best capture the distribution

of states.

We conducted this analysis and observed global agreement between the results
of both of our datasets. As we observed substantial particle damage in our Vitrobot-
prepared dataset, we thought it unwise to utilize that dataset for in-depth analysis and
therefore, we focus on the larger, and higher-resolution dataset collected at NCCAT
below. For analysis, the entire particle set, prior to 3D classification and consisting of
~170k particles, was utilized. The particle alignments from the same C1 consensus
refinement were used as input for each of the three methods as current methods to
investigate particle heterogeneity are highly dependent upon initial pose assignment of
individual particles. Globally, all three methods agreed with our ensemble structures, in
that the dominant source of structural variability within our particle population could be
described by the open-closed transition of individual domains, and that sACE-N displayed
more variability than sACE-C.

Initially, we performed 3D variability analysis (3DVA) in cryoSPARC to provide a
direct comparison to the previously published analysis of the partial ACE cryo-EM
structure (57) (Figure 6A). Trajectories were reconstructed along the top 5 principal
components of variance (Movie S9). Surprisingly, the trajectory describing the variance
along the first principal component vector suggests a dominant interdomain motion,
where the sACE-N/N region bends towards the sACE-C/C region, similar to what we
observed for trajectories along the top principal component vectors in analysis of our MD
simulations (Movie S7,8), yet this variance is not reflected in our ensemble structures.
The trajectories along the remaining principal component vectors comprise varying
combinations of individual domains undergoing the open-close transition. Consistent with
our previous observations, the range of presumptive motion in these trajectories is greater
in sACE-N than sACE-C. Previous 3DVA performed on the sACE monomer structure
revealed 4 types of potential motion, termed bending, pivoting, breathing, and jumping
(33). The breathing motion matches the open-close transition that dominates most of our
principal component vectors, while the bending motion appears to correlate with the
variance described by our top principal component vector. We did not observe any
potential motions reminiscent of the jumping or pivoting motions in our 3DVA.
Furthermore, the magnitude of the bending motion in our 3DVA is substantially less than
previously reported for the partial dimeric sACE structure, which likely suffered from
substantial denaturation (33). Together, this highlights a significant difference in the

conformational dynamics between our sACE dimer and previously reported cryo-EM
structures. While there is disagreement between the 3DVA results for the two datasets, it
should be noted that our MD simulations display a greater range of motion than either
cryo-EM dataset offers, and the varieties of motion described previously are clearly
observed within the PCA trajectories, especially when sACE is glycosylated (Movie S8).
As mentioned above, this type of cryo-EM analysis is sensitive to artifacts introduced by
particle damage and denaturation, underscoring the importance of preserving protein
integrity via sample preparation and the use of Chameleon to reduce the exposure to air-
water interface. This also suggests dimerization functions as a key constraint to the
conformational dynamics of sACE.

We then moved on to analyze our data using cryoDRGN, which revealed that the
data formed a single cluster across multiple dimensions (Figure 6B, S14). Trajectories
generated along five principal component vectors of structural variance were dominated
by the open-close transition, primarily in sACE-N (Movie S10). We observed little variance
in sACE-C and, unlike 3DVA, cryoDRGN did not reveal any substantial interdomain
motions. Faced with this discrepancy between the 3DVA and cryoDRGN results, we
employed a third method, RECOVAR, to further analyze our data. Similar to cryoDRGN,
RECOVAR revealed that our data formed a singular cluster across multiple dimensions
of analysis. Volumes representative of kmeans cluster reconstructions were similarly
consistent with our cryoDRGN results. RECOVAR indicated that the primary source of
structural variance could be explained by the open-close transition, primarily of the sACE-
N domains. With this in mind, we repeated the RECOVAR analysis with a focus mask
around the sACE-N/N region. We then estimated the conformational density by
deconvoluting the multidimensional latent space with a two-dimensional principal
component analysis. The resulting density plot revealed four clusters which corresponded
to different conformations of the sACE-N/N region (Figure 6C). The most populous cluster
corresponded to a state where both sACE-N subunits adopted an open conformation.
The second largest cluster, and the smallest cluster, corresponded to a state where one
sACE-N was open and the other was closed. The discrepancy in population size between
these clusters is likely due to bias in the initial particle orientation, rather than a subunit-
specific preference for the open state. As the O/C state and the C/O state are 180 degree
rotations of each other, particle assignment to either cluster is likely influenced by the

initial particle orientation of the complete dimer, and we currently lack the data to discern
any allosteric implication to the orientation assignment. The final cluster corresponded to
a state where both sACE-N subunits adopted the closed conformation. We then used
RECOVAR to generate a trajectory through the latent space to predict a potential
transition pathway from the closed-closed state to the open-open state (Movie S11). This
trajectory revealed that one subunit would open, followed by the second, and raised the
question of potential allostery or coordinated motion between the adjacent sACE-N
domains.

To investigate coordination between the sACE-N domains, we performed 3D
classification in cryoSPARC with a focus mask on a single sACE-N domain and requested
2 classes (Figure 6D). The particles were split roughly evenly between the two classes,
one resulting in an open state and the other being closed. We then used these classes
for another round of 3D classification with a focus mask on the opposite sACE-N domain
and again asked for 2 classes. Again, these particles were split roughly evenly among the
classes, leaving us with four similarly sized classes adopting the same conformations as
the clusters revealed by RECOVAR.

Discussion

Our cryo-EM and MD simulations analysis of the extracellular regions of dimeric
ACE (sACE) offers insights into the stability of sACE catalytic domains. We show that the
sACE dimer interface consists of three major components, protein-protein interfaces
between sACE-N, glycan-glycan interaction between sACE-N, and protein-protein
interface between sACE-C. The interface between sACE-N, in general, is 50% larger and
has higher numbers of hydrogen bonds and salt bridges than that between sACE-C. The
necessity for the stronger interaction between sACE-N is likely due to the fact that sACE-
N is distal to the transmembrane helix so that the additional reinforcement is required.
The weaker interaction and lack of contribution from glycans in the interface between
sACE-C could also contribute to the observed preferential denaturation of sACE-C
despite the high structural similarity between sACE-N and sACE-C. The denaturation is
likely due to the exposure to air-water interface during vitrification, a common issue in
single particle cryo-EM analysis (40, 41). We have partially overcome the denaturation
issue of sACE-C by optimizing the buffer pH to increase the melting temperature and
using the faster plunging time during vitrification afforded by Chameleon to reduce the

exposure to air-water interface (Figure 1). However, we did not completely prevent the
preferential denature of sACE-C during vitrification.

ACE is part of a family of proteases capable of effectively degrading Aβ. This family
includes M16 metalloproteases like IDE and PreP, and M13 metalloproteases such as
neprilysin and endothelin converting enzyme. Integrative structural analysis has
demonstrated how IDE employs conformational dynamics, charge, and secondary
structure complementarity to selectively capture peptide substrates into its sizable
catalytic chamber, subsequently unfolding and degrading amyloidogenic peptides such
as insulin and Aβ, without degrading structurally similar yet non-amyloidogenic peptides
(30, 31, 58). The other mentioned enzymes also belong to this group of chamber-
containing proteases, known as cryptidases (29, 30, 59). Similar to cryptidases, structural
analysis reveals that ACE must undergo an open-closed transition to capture, engulf, and
degrade its oligopeptide substrates, such as angiotensin 1 and bradykinin.

Our apo-sACE structures and MD simulations identify a fulcrum mechanism that
governs the open-closed transition of sACE-N, while the interplay between sACE-N and
the D3a domain likely plays a crucial role in the corresponding transition in sACE-C.
Analogous to how the binding of MB60, a PreP inhibitor, to the hydrophobic pocket
triggers the open to closed transition, factors such as binding of substrates and ACE
inhibitors may alter the balance, leading to the regulation of ACE activity (29). We also
observe that the catalytic pocket of each sACE catalytic domain is not large enough to
engulf the entire Aβ. Additionally, the distances between the domains (sACE-N to sACE-
C, sACE-N to sACE-N, or sACE-C to sACE-C) are too large for these domains to bind Aβ
simultaneously (60). Therefore, ACE likely employs a different mechanism from IDE and
PreP to selectively bind and degrade Aβ. Future integrative structural studies await to
reveal how ACE selectively recognizes and degrades a diverse array of substrates.

Globally, we observed that sACE-N is both more open and more dynamic than
sACE-C in the absence of substrate. We speculate that the increased dynamicism of
sACE-N may be due to a weakening of the hydrophobic latch region relative to sACE-C,
which renders the closed state more thermodynamically favorable in sACE-C, along with
increased stabilization of the sACE-N open state through the fulcrum mechanism
described above. These factors result in sACE-N demonstrating a preference for the open
state, while sACE-C appears to prefer the closed state, in the absence of substrate. The

preference of sACE-C towards the closed state may explain why it displays a greater

catalytic activity compared to sACE-N for substrates such as angiotensin-I (32). The

difference in dynamics between sACE-N and sACE-C also offers new pathways forward

in the development of domain-specific inhibitors to better modulate ACE function. The

fulcrum-associated region centered on the K73-D189 interaction stands out as a

promising new druggable target to modulate the openness of sACE-N independent of

sACE-C. Additionally, given the range in size of ACE substrates, i.e. amyloid β (42

residues) vs bradykinin/angiotensin-I (9/10 residues), it is likely that the openness of

individual ACE domains functions as a key mediating step in substrate recognition and

altering the open-to-close dynamics of ACE domains may prove an effective strategy to

modulate ACE activity towards specific subsets of substrates. These insights come with

the caveat that they are derived from the apo-state dynamics of sACE and it remains

unknown how sACE dynamics are influenced by the presence of substrate. Studies in the

presence of substrate will be especially vital in exploring potential mechanisms of ACE

allostery.

The interplay between ACE-N and ACE-C within the ACE dimer has complicated

consequences for ACE function, showing negative cooperativity for enzymatic activities

while exhibiting positive synergy in Aβ cleavage by ACE-N and activity at the bradykinin

receptor (13, 36-39). Understanding the conformational dynamics in both inter- and

intradomain motions of the extracellular regions of the ACE dimer holds promise for

designing ACE modulators to improve and broaden the therapeutic use of ACE

modulation. Our cryo-EM heterogeneity analysis of sACE alone has revealed coordinated

yet complicated patten of open-closed transitions across four catalytic domains using

cryoDRGN and cryoSPARC 3DVA (Figure 6A,B, Movies S9,10). RECOVAR analysis

further illustrates the distribution of open-closed states within the sACE-N dimer and the

potential pathways for such transitions (Figure 6C). Additionally, focused classification

using cryoSPARC suggests no bias in the conformational state of sACE-N within a dimer,

irrespective of the state of its counterpart (Figure 6D). Altogether, these findings establish

a framework for understanding how the complex open-closed transitions within the four

catalytic domains of the sACE dimer contribute to its allosteric behavior (13, 36-39).

3DVA, cryoDRGN, and RECOVAR all appeared to display varying degrees of

coordinated motion between sACE domains. However, this type of analysis lacks an

explicit time component, so it is not possible to draw concrete conclusions regarding

coordinated motion. Likewise, while MD simulations contain an explicit time factor, most

of our MD simulations closed quickly, and did not reopen. While we observed an apparent

association between intradomain and interdomain motion in our PCA (movies S7, S8),

the analysis removes time dependency. Furthermore, the aforementioned allostery of

ACE was observed under the condition where substrate is present (13, 36, 38).

Complicating matters further, this conformational heterogeneity appears to stem partially

from subtle yet significant conformational differences between the two chains within an

ACE dimer and the inability of cryoSPARC to accurately align these conformationally

distinct chains, likely due to particle pseudo-symmetry (Figure S15). Furthermore,

CryoDRGN and RECOVAR revealed smaller magnitudes of potential interdomain

motions than cryoSPARC 3DVA and they were often accompanied by various

combinations of individual domains opening and/or closing. This suggests that the

apparent discrepancy in the magnitude of inferred interdomain motion could be a result

of the different theoretical approaches of dimensionality reduction and trajectory

generation employed by the respective methods of analysis and underscores the

importance of incorporating multiple methods of analysis in the absence of a singular

"gold-standard" of heterogeneity analysis. Conversely, it is possible that interpretation of

any potential interdomain motion is complicated by the preferential denaturation of sACE-

C. Principal component analysis of our MD simulations revealed the dominant source of

motion to be alterations in the position of sACE-C/C relative to sACE-N/N. We speculate

that many of these relative orientations cannot tolerate the rigors of the cryo-EM grid-

making process and may be the source of the observed preferential denaturation of

sACE-C. Future studies will need to focus on cryo-EM analysis of ACE in the presence of

substrates, ideally using data in which partial denaturation is substantially minimized.

Single particle cryo-EM holds promise for revealing the conformational dynamics

of biological macromolecules from embedded heterogeneity. Consequently, cryo-EM

heterogeneity analysis is an active and rapidly evolving field providing insights into the

conformational dynamics and allostery of macromolecules (53). However, many

challenges exist in using cryo-EM heterogeneity analysis due, in part, to the low signal-

to-noise ratio in observed micrographs. Often, these analyses address questions difficult

or impossible to tackle experimentally, leaving no ground truth for validation. We

encounter such challenges in our heterogeneity analysis as well. Consistent with our single particle cryo-EM structural analysis, we observed the expected intradomain open-closed transition using cryoDRGN, cryoSPARC 3DVA, RECOVAR, and all-atom MD simulation. Surprisingly, we did not observe such motion using cryoSPARC 3DFlex, a neural network-based method analyzing our cryo-EM data of sACE (54). Central to the working of cryoSPARC 3DFlex is the generation of a tetrahedral mesh used to calculate deformations within the particle population. Proper generation of the mesh is critical for obtaining useful results and must often be determined empirically. Despite several attempts, we were unable to obtain results from 3DFlex comparable to what we observed with our other methods. Even using the results from our 3DVA as prior input to 3DFlex, the largest conformational change we observed was a slight wiggling at the bottom of the D3a subdomain (Movie S12). The authors of 3DFlex note that 3DFlex struggles to model intricate motions, and the implementation of custom tetrahedral meshes currently requires a non-cyclical fusion strategy between mesh segments. Given these limitations, and the complexity of sACE conformational dynamics, it appears that sACE, as a system, is not well-suited to analysis via 3DFlex in its current implementation. Similarly, MDSPACE holds tremendous promise as a method for investigating conformational dynamics from cryo-EM data (61). MDSPACE integrates cryo-EM particle data with short MD simulations to fit atomic models into each particle image through an iterative process which extracts dynamic information. However, the lack of GPU-enabled processing for MDSPACE requires either a dedicated a computational setup that diverges from most other cryo-EM software, or access to a CPU-based supercomputer, which severely limits the accessibility of such software. Despite these challenges, both 3DFlex and MDSPACE use promising approaches to study protein conformational dynamics. We look forward to exploring effective methods to incorporate these strategies into our future research.

Advances in single-particle cryo-EM analysis have enabled detailed structural examination of the sACE dimer. Beyond its implications for future drug discovery efforts aimed at enhancing ACE-based therapies, sACE can serve as a valuable model of a multi-domain glycoprotein for studying methods to preserve sample integrity, particularly due to the high propensity of sACE-C to denature during vitrification. Furthermore, considering the aforementioned challenges and caveats, we propose that, due to its size,

biomedical importance, and the complexity of its motions, sACE also represents an excellent model system for evaluating and improving cryo-EM heterogeneity analysis.

Experimental procedures

Expression of human sACE: Human 293T and 293F cells were maintained in DMEM + 10% fetal calf serum (FCS). The lentiviral plasmid for the expression of the extracellular domain of human ACE (sACE, aa 1-1235) that contains a 6-histidine tag at its C-terminus was made by Vectorbuilder and co-transfected with helper plasmids to human 293T cells for lentiviral production. Human 293F cells were infected with the resulting virus and selected for stable sACE expression. 293F that stably expressed sACE grew poorly in Free-style 293 media (ThermoFisher 12338018). Thus, for ACE production, 293F cells that stably expressed sACE were first grown in DMEM+10% FCS at 37°C and gradually shifted into Free-style 293 media by reducing FCS by first in Free-style 293 media + 2% FCS and then to media without FCS. These cells were then shifted to and maintained in 30°C for 3-5 days for ACE production to reduce proteolysis.

Purification of soluble human ACE: Culture supernatant was centrifuged at 3,000xg for 5 min, diluted 10-fold with 25 mM Tris-HCl pH 8, filtered, and run over a Source Q column pre-equilibrated with 25 mM Tris-HCl pH 8. Column was washed to baseline with Tris-HCl pH 8 and bound protein was eluted with a linear gradient of NaCl to 1 M. sACE containing fractions were pooled and EDTA was added to 10 mM to strip the catalytic zinc ion. Sample was concentrated and run over a Superdex S200 SEC column pre-equilibrated with the desired cryo-EM buffer.

Differential scanning fluorometry (DSF): To optimize the grid making conditions, differential scanning fluorimetry was applied to screen 96 buffers conditions. DSF was performed with 3 mg/ml sACE and 10X Sypro Orange in 20 μl buffers in a Thermo Fisher Step ONE RT-PCR thermocycler. Using the melting temperature (Tm) and slope as selective criteria, 25 mM citrate pH 5.5, 25 mM sodium phosphate pH 5.5, 150 mM NaCl, 10 mM EDTA was identified as the optimal condition for the highest Tm and optimal slope, which was used for all subsequent cryo-EM experiments.

Cryo-EM data collection and processing: Peak fractions were collected off the SEC column and used to immediately prepare grids, resulting in a protein concentration of ~4.5 mg/ml. Quantifoil R1.2/1.3 Cu200 grids were glow-discharged in air and grids were

prepared by Vitrobot mark 3. 3.5 μl of sample was applied to the grid in the chamber with

100% humidity, 25°C. Excess sample was blotted for 1 sec, blot force 2, and grids were

plunge frozen in liquid ethane. All images were acquired using a Titan Krios microscope

(FEI) operated at 300KeV with a Gatan K3 direct electron detector. A dataset of ~3,600

micrographs was collected at The University of Chicago Advanced Electron Microscopy

facility. The dataset was processed using cryoSPARC (Figure S2) (62). Briefly, a four-

domain sACE dimer model was generated using sACE-N dimer (PDB 7Q4D) and sACE

monomer (PDB 7Q3Y) and adjusting the distance between ACE-C domains to avoid

steric clashes as a template for particle picking (33). After template picking, 1.1 million

particles were extracted with a box size of 360 pixels. Several rounds of 2D classification

were used to remove junk particles. ~577K particles were processed in successive rounds

of *ab initio* reconstruction and heterogeneous refinement. sACE grids were also prepared

using Chameleon and collected a dataset of ~19,000 micrographs at the National Center

for Cryo-EM Access and Training. For those, Quantifoil Active R 1.6/0.9 Cu/Rh 300 were

coated with gold using a Safematic CCU-010 Compact Coating Unit. Briefly, grids were

placed nanowire side down on a glass slide, gold disks were sputtered with a 35 nm target

thickness at 1.5 x 10^{-2} Torr, and a process current of 30 mA. Grids were glow discharged

with air at 12mA for 80 secs using the internal Chameleon glow discharger. The grids

were plunged at 160 to 180 milliseconds. All images were acquired using a Titan Krios

microscope (FEI) operated at 300KeV with a Gatan K3 direct electron detector (Gatan).

Images were automatically acquired using Leginon (63) using collection parameters as

shown in Table S2. This dataset was processed in cryoSPARC (Figure S3). Briefly,

templates were generated from our full-length ACE structure described above using

vitrobot-prepared grids. Micrographs with a CTF fit resolution worse than 4Å were

removed from the dataset, leaving ~12,000 micrographs for template picking, which

yielded 10.7 million particles. The particle set was trimmed to ~2.8 million based on quality

metrics, particles were extracted with a box size of 256 pixels and run through 2 rounds

of 2D classification, reducing the particle number to ~660,000. 2D classes were

dominated by 4 domain views, and yielded a much greater diversity of views, indicating

that Chameleon preparation produced a dramatic improvement in particle quality. Initial

3D homogeneous refinement reached 3.6 Å, matching the previous final structure. Four

maps from 3D classes that contains all four catalytic domains of ACE, one from particles

derived from grids prepared by using vitrobot and three from those prepared by using

Chameleon, were used for structure building using Chimera and Coot and the structure
was refined using Phenix real space refinement. Structural analysis was performed using
PYMOL and Chimera.

Heterogeneity analysis: All heterogeneity analysis was performed on the full final
particle set, encompassing all of the NCCAT structures. Initially, cryoSPARC 3DVA was
performed on both the NCCAT- and UChicago-collected datasets and results were found
to be consistent, so the NCCAT-collected dataset was used for further analysis, given its
higher particle number and quality. Input particle poses for each method of heterogeneity
analysis were derived from the same C1 refinement job in cryoSPARC, which reached
2.88 Å. 3DVA was used to solve the top 5 modes of variance with a filter resolution of 6
Å. Particles were downsampled to a box size of 128 pixels in cryoDRGN to speed up
processing. The same downsampled particle set was used for both cryoDRGN and
RECOVAR. CryoDRGN was run with default parameters and the analysis was performed
on the top 5 principal components for direct comparison to cryoSPARC 3DVA. RECOVAR
pipeline.py was initially run on the full sACE structure. Reconstructions representative of
kmeans cluster centers revealed the dominant source of structural variability to be the
open-close transition in sACE-N. As a result, RECOVAR was run again with a focus mask
on sACE-N/N and a zdim=4. Conformational density was estimated with a 2-dimensional
PCA, and an index value of 3 was chosen for further analysis. The points in latent space
representing the top 12 stable conformational states were estimated from the top 20% of
particles and representative volumes were generated. The generated states populated
four clusters and the 12 states were simplified to 4 distinct conformations where sACE-
N/N adopted the open-open (OO), open-closed (OC), closed-open (CO), and closed-
closed (CC) states (Figure 6C). We then generated a trajectory through the latent space
using the OO and CC representative volumes as endpoints (Movie S11).

All-atom MD simulations and analysis: The sACE-3.65 structure was used to initiate
all MD simulations. For the glycosylated sACE simulations, N-linked glycans were built
onto all 12 known N-linked glycosylation sites (residues 9, 25, 45, 82, 117, 289, 416, 480,
648, 666, 685, 731) with the complex glycan BGLC-BGLC-BMAN-AMAN-BGLC-ANE5-
AMAN-BGLC-BGAL-ANE5-AFUC using CHARMM-GUI (64). Simulations were prepared
in QwikMD or CHARMM-GUI and run at pH 7, 150 mM NaCl, 310 K, 1 atm with periodic
boundary conditions in NAMD3.0 using the University of Chicago Beagle3 partition (65).

Explicit solvent was described with the TIP3P model (66). Data was analyzed in VMD

1.9.4 (67) using a combination of built-in plugins and custom TCL scripts, as well as

custom Python scripts. Analysis scripts are available upon reasonable request.

Figure + figure legends:

Figure 1: Cryo-EM analysis of human sACE dimer. (A) Diagram for the key features of domain on primary sequence of human ACE. The construct used in this study contains the first 1235 residues, which we refer to as the soluble region of ACE (sACE) however, the first 29 residues comprise a signal. By convention, sACE is labeled based on the mature, processed peptide (68). D1a and D3a domains, also called "lid" encompasses residues 1-98 and aa 616-696, respectively. D1b and D3b domains each have two discrete segments; residues 263-436 and 496-574 for D1b and residues 868-1031 and 1091-1171 for D3b. D2 encompasses three discrete segments, residues 99-262, 437-495, and 575-615 while D4 domain encompasses residues 697-867, 1032-1090, and 1172-1202. sp = signal peptide. asterisk = zinc binding motif. Colored by sub-domain: D1a blue, D1b gold, D2 magenta, D3a cyan, D3b green, D4 purple. (B) 2D classification of human sACE particles from grids made by vitrobot (360 pixels box size) and Chameleon (256 pixels box size). Clear four domain classes visible in the Chameleon-derived classification are boxed in red, similar views are lacking in the vitrobot dataset. White scale bar measures 100 Å. (C) Full-length sACE 3D volumes, colored by sub-domain as in (A). Chain B sub-domains are depicted as lighter tones of their chain A counterparts. Glycan density is shown in gray. See Figure S2 for vitrobot-prepared data processing details, Figure S4 for Chameleon-prepared data processing details and Table S2 for data refinement statistics.

Figure 2: Overall structure of human sACE. (A) Overlay comparing sACE-N states highlighting the structure differences between the closed "C", intermediate "I", and open "O" states. (B) Overlay comparing sACE-C states highlighting the structure differences between the closed "C", and intermediate "I" states. We define the state based on the distance between the edge of the D2/4 domain bordering the catalytic cleft (residues 121-126 or 721-726) and the tip of the D1/3a region (residues 41-51 or 647-657): closed <15 Å, intermediate >15 Å and <19 Å, open >19 Å. (C) Overall dimer comparisons. Black bracket depicts the D1/3a-D2/4 distance measurement used to define the state of each domain, as described above. (D) Table of openness measurements for each domain per structure.

Figure 3: sACE dimerization interfaces. (A) Overlay comparing sACE-N/N interface (blue)

and sACE-C/C (gold) interfaces. Interfaces adopt the same secondary structure but

interacting residues vary between them. (B) Residue-specific interactions at the sACE-

N/N interface, see text for details. (C) Unsharpened Coulomb potential density map (blue)

showing density corresponding to glycan-glycan interaction from N82 as part of sACE-

N/N interface. Sharpened map is shown in magenta for reference. (D) Residue-specific

interactions in the sACE-C/C interface, see text for details.

Figure 4: Comparison of molecular dynamics simulations with cryo-EM data. (A) Alpha carbon displacement values for each residue were calculated by comparing each of our cryo-EM structures against one another in a pairwise manner and averaged. Individual residues are colored by sub-domain as defined in figure 1. Alpha carbon displacement values were also mapped onto the structure of sACE to better visualize the mobile regions. Residues are colored by degree of displace from blue (no displacement) to red (high displacement, values in Angstrom, as indicated). Numbers denote regions of interest: 1, the top of the D1a sub-domain, 2, the bottom of the D1a sub-domain, 3, a flexible loop region within the D1b that interacts with the bottom of the D1a sub-domain near the catalytic cleft, 4, the bottom of the D3A sub-domain. (B) Alpha carbon RMSF values for a representative 100 ns subset from our non-glycosylated MD simulations that demonstrates the simulated dynamics correlate well with displacement calculated from the cryo-EM structures. (C) Alpha carbon RMSF values for the entirety of our non-glycosylated MD simulations. Mobile regions agree with predictions from the cryo-EM structures, yet the magnitude is dampened due to the protein rapidly transitioning from an open state to a closed state and remaining closed for most of the simulation time. (D) Alpha carbon RMSF values for a representative 100 ns subset from our glycosylated MD

Figure 5: Structural mechanism of the sACE open/close transition. (A) sACE-N overlay
comparing the open (left panel) and closed (right panel) states in detail. The open state
is stabilized by interaction between residues in the D1a (blue) and D2 (magenta) regions
that, notably K73-D189. In the closed state, the K73-D189 interaction is broken. (B)
Overlay of the sACE-N open (dark shades, sub-domains colored as above) and closed
(light shades, sub-domains colored as above) states showing the range of motion. The
D1a region rotates about a fulcrum region described in (A), while the D1b region moves
as a rigid body. Front view arrows depict the "fulcrum" motion of the D1a subdomain, with
the top and bottom of the subdomain moving in opposite directions. Side view arrows
depict rigid body motion of the D1a and D1b subdomains moving together. (C) Overlay of
sACE-C closed (light shades, sub-domains colored as above) and intermediate (dark
shades, sub-domains colored as above) states. Unlike sACE-N, the "top" of the D3a
region is constrained by its connection to sACE-N and largely immobile. The primary
source of opening is only the motion of the D3a tip. We did not observe any open state

structures of sACE-C, suggesting a smaller range of motion relative to sACE-N. Front

view arrow depicts motion of D3a subdomain. Only the bottom of the subdomain moves,

in contrast to the "fulcrum" motion observed in the D1a subdomain. Side view depicts the

rigid body motion of the D3a and D3b subdomains moving together. (D) Comparison of

the hydrophobic "latch" region formed in the closed state between residues of the D1/3a,

D1/3b, and D2/D4 domains. V724 in sACE-C has been replaced by T124 in sACE-N,

suggesting that the closed state in sACE-N may be less stabilized than the sACE-C

closed state. (E) Example all-atom MD simulation tracking the openness of one sACE-N

region (black line) and this distance between K73 and D189 (red line). These residues

form a salt bridge early in the simulation when sACE-N is open (left inset) but the

interaction breaks as sACE-N transitions to the closed state (right inset). Distance

measurements for MD simulations were consistently greater than distance values in our

static structures and cannot be directly compared to Figure 2.

Figure 6: Cryo-EM heterogeneity analysis. (A) Visualization of the structural changes revealed by cryoSPARC 3DVA trajectories calculated along two principal components (PCs) of structural variance. Starting states are showing in cyan, ending states in gray. PC 0 reveals a large, inter-domain bending motion accompanied by the open/close transition in sACE-C. PC 1 and the remaining PCs are dominated by the open/close transition of individual regions. See Table S3, and Movie S9 for additional details. Arrows depict generalized motions. (B) Visualization of the structural changes revealed by the cryoDRGN trajectories calculated along two PCs of structural variance. Starting states are shown in cyan, ending states in gray. PCs are dominated by the open/close transition of individual regions. See Table S3, and Movie S10 for additional details. Arrows depict

generalized motions. (C) Analysis in RECOVAR with a focus mask on sACE-N/N reveals

that particles adopt roughly four clusters within the latent space (heat map of particle

density) corresponding to the open-open (OO, white square), open-closed (OC, white

dot), closed-open (CO, white dot) and closed-closed (CC, white star) states of sACE-N/N.

A trajectory estimating the path through latent space corresponding to the structural

transition from sACE-N/N CC state to the OO state (blue points) suggests that individual

sACE regions transition at different rates, as indicated by the size of the transition arrows

between states. See Movie S11 for trajectory. (D) Focused 3D classification was

performed in cryoSPARC to explore evidence of coordinated motion between sACE-N

regions. 3D classification focusing on one sACE-N region revealed 2 roughly equal

classes of particles: open and closed. Subsequent 3D classification focused on the other

sACE-N region again revealed 2 roughly equal classes, suggesting the lack of

coordinated motion between sACE-N regions in the absence of substrate.

Data availability:

Unprocessed micrographs, particle stacks, and associated pose/CTF information have been deposited to EMPIAR under the accession numbers: EMPIAR-12181 (Vitrobot-prepared grids) and EMPIAR-12484 (Chameleon-prepared grids)

EM structures and maps have been deposited to the EMDB and PDB, respectively, under the following accession numbers:

sACE-2.99: 9D5S and EMD-46581

sACE-3.05: 9D5M and EMD-46579

sACE-3.15: 9D55 and EMD-46574

sACE-3.65: 9CLX and EMD-45733

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Supplemental figures

Figure S1: DSF optimization of sACE grid-making conditions. 48 buffer conditions were screened for sACE stability in the presence and absence of EDTA (see table S1). Citrate-phosphate buffer pH 5.5 (red) was chosen for the best combination of stability (temperature shift right) and homogeneity (steepness of transition slope). Two other citrate-phosphate conditions, pH 5 and pH 4.5 gave better stability, but a more gradual transition slope, and so were not chosen.

Figure S2: Processing workflow for dataset using Vitrobot-prepared grids. 3,653 micrographs were collected on a 300 keV Titan Krios at the University of Chicago. A full-length model of sACE was manually generated by combining available PDB structures of ACE-N and ACE-C. The resulting PDB file was converted to an EM map in Chimera and imported into cryoSPARC to generate particle picking templates. Using these templates, 1,171,502 particles were picked and extracted. Successive rounds of 2D classification were used to trim the particle number down to 577,123 particles. *Ab initio* reconstruction followed by heterogeneous refinement yielded a full sACE dimer map (214,893 particles), a map representing only the N/N dimer (268,471 particles), and 93,759 particles that mapped to an uninterpretable junk class. The particles mapping to the full length sACE structure were then transferred to PCA-based 3D classification. 103,507 particles mapped to low resolution partial structures that could not clearly be assigned structural significance and were discarded as junk. Of the remaining particles, 20,533 particles mapped to a three-domain sACE structure. The remaining 88,930 particles mapped to a

full-length sACE structure that was refined to 3.65 Å (sACE-3.65). Scale bar in
representative micrograph shows 200 Å, scale bars in 2D classes show 100 Å.

Figure S3: Assessment of map quality from Vitrobot prepared grids. (A) Angular
distribution. (B) Local resolution colored as blue-cyan-green-yellow-red spectrum from 3-
5 Å. (C) Example residue fits to density map. (D) 3D-FSC results.

Figure S4: Processing workflow for dataset using Chameleon-prepared grids. 18,281
micrographs were collected at the National Center for CryoEM Access and Training. After
motion correction and CTF estimation, the dataset was trimmed to include only
micrographs with a max CTF value lower than 4 Angstroms. The sACE-3.65 structure
was used to generate 2D templates for particle picking. 10,708,232 particles were
originally picked. This number was trimmed to 2,580,990 based on quality metrics and
visual observation. The resulting particles were used in successive rounds of 2D
classification to eliminate junk particles, reducing the particle count to 662,219. *Ab initio*
reconstruction followed by heterogeneous refinement revealed four classes: 165,395
particles mapped to a class we interpret as the sACE-C/C dimer, 140,038 particles
mapped to a class we interpret as the sACE-N/N dimer, 50,647 particles mapped to a 3-
domain sACE class, and the remaining 306,139 particles mapped to a full-length sACE
class. PCA-based 3D classification was used to eliminate junk particles, and the
remaining ~170k particles were refined with C2 symmetry to 2.8 Angstroms. The D1
domains in the C2 reconstruction displayed poor density, likely resulting from molecular
motion, so a final round of 3D classification and refinement with C1 symmetry was
performed to finalize three states of full-length sACE. Scale bar in representative
micrograph shows 150 Å, scale bar in 2D classes show 100 Å.

1010

Figure S5: Assessment of map quality from Chameleon prepared grids. Angular

distribution, local resolution, example density fit and 3D-FSC curves for (A) sACE-3.05,

(B) sACE-3.15, and (C) sACE-2.99.

Figure S6: B factor comparison of all structures. Full-length sACE structures colored by
B factor. Scale ranges from a B factor of 10 (dark blue) to 225 (red). Protein is depicted
as cartoon, glycans are shown as sticks.

1018

Figure S7: sACE-N/N dimers compared to previously published structures. Previously
identified sACE-N/N dimers 2XYD (green, which is similar to structures of following pdb
codes: 3nxq, 4bxk, 4bzs, 4ca6, 4ufa, 6f9r, 6f9v, 6h5x, 6qs1, 6tt1, 6tt3, 6tt4, 7q25, 7q26,
7q27, 7q29, 7z6z), 2C6F (cyan, which is similar to structures of following pdb codes:
2c6n, 5amb, 5amc), 4UFB (magenta, which is similar to structures of following pdb codes:
5am8, 5am9, 5ama, 6en5, 6en6, 6zpq, 6zpt), and 7Q4D (orange) aligned to sACE-2.99
(gray).

1026

1027 Figure S8: sACE exhibits a continuous gradient of structural heterogeneity. All 16
1028 individual domains from our sACE structures were aligned to their respective D2/4
1029 regions, revealing a continues gradient of "openness" (gray) from the most closed
1030 conformation (green) to the most open conformation (magenta).

1031

Figure S9: RMSD analysis of non-glycosylated MD simulations. RMSD values were
calculated for the D1/3a (red), D1/3b (gray), and D2/4 (orange) regions and compared to
the RMSD values for the global domain (blue) to identify the primary source of structural
variance in the non-glycosylated sACE MD simulations. RMSD was calculated using
backbone atoms only. For each panel, the Y-axis is RMSD value, the X-axis is simulation
frame. Every frame represents a time step of 0.1 ns.

1038

Figure S10: Principal Component Analysis. PCA was performed using the alpha carbon
coordinates for every residue in every frame of our MD simulations along with each of our
cryo-EM structures. Red X's denote data from cryo-EM structures. MD simulation data is
colored by run as indicated. Non-glycosylated MD simulation data is shown on the left,
glycosylated MD simulation data is shown on the right. The conformational space
explored in our MD simulations is significantly greater than the range of conformations
observed in our cryo-EM structures, as expected. The most significant source of
conformational variation in our MD simulations relative to our cryo-EM structures is the
reposition of sACE-C/C relative to sACE-N/N (Movies S7,8).

Non-glycosylated
MD simulations

Glycosylated MD
simulations

1048

Figure S11: Clustering Analysis. The conformational space sampled by each domain in

our non-glycosylated (left) and glycosylated (right) MD simulations was assessed by

tracking the angle formed by the centers of mass of the D2-D1a-D1b (or D4-D3a-D3b for

sACE-C) relative to the openness of the domain (D1/3a-D2/4 measurement as shown in

figure 2) for each frame of our simulations and plotted as density. The corresponding

measurements from our cryo-EM structures are shown in red.

Non-glycosylated   Glycosylated

1055

1056 Figure S12: RMSF comparison of glycosylated vs non-glycosylated MD simulations.
1057 RMSF values for each residue over the course of all glycosylated and non-glycosylated
1058 MD simulations. The RMSF values for each residue in the non-glycosylated simulations
1059 were subtracted from the values for each residue in the glycosylated simulations and
1060 plotted. Chain A residues are shown in blue, chain B residues are shown in orange.

1061

1062 Figure S13: Shielding of glycans in MD simulations. 200 representative frames of glycan
1063 orientations (light blue) were overlaid the sACE structure (pink shade by chain) to
1064 visualize the amount of protein surface insulated from solvent access by glycans.

1065

1066 Figure S14: CryoDRGN clustering of particles. Top panel shows the deconvolution of the
1067 latent space along specified principal component vectors and the path of each trajectory
1068 generated as movies (see Movie S10). Bottom panels show particle clustering resulting
1069 from UMAP deconvolution of the latent space with the path of each trajectory generated.

1070

Figure S15: Artificially observed interdomain motion caused by the subtle yet noticeable
differences between the two chains within the sACE dimer and the choice of chain pairs
between sACE dimer structures. A. Overall structure of the sACE dimer and domain
assignment. B. Ribbon representation of sACE-2.99 structures where the assignment of
chain A and chain B within the sACE dimer is swapped. C. Ribbon representation of the
sACE-3.05 structure. D. Alignment of the sACE-2.99 structure with sACE-3.05. When
sACE-2.99 A/B chains are aligned with sACE-3.05 A/B chains, one domain is well-aligned
but the other three are not. There is an ~5° rotation of D4/D4 relative to D2/D2 from the
alignment. E. The alignment of two structures that are the same as D except chain A and
chain B in sACE-2.99 is swapped. When sACE-2.99 B/A chains are aligned with sACE-
3.05 A/B chains, three domains are well-aligned and the remaining one depicts the open-
closed transition of sACE catalytic domain. In this case, D2/D2/D4/D4 of the two
structures are well aligned and there is no rotation.

1084

Supplemental tables

1086   Table S1: List of buffers used in DSF optimization of grid-making conditions.

1087   Table S2: CryoEM processing statistics for both datasets

1088   Table S3: Summary of MD simulations

1089   Table S4: Summary of heterogeneity analysis

Supplemental movies

Movie S1: Overview of sACE-3.65 structure. Coulomb potential density map colored by
subdomain as Figure 1, followed by cartoon-depicted model of sACE-3.65 structure
(colored by sub-domain) fit into its Coulomb potential map (gray), and then cartoon-
depicted model of sACE-3.65 structure (colored by sub-domain).

Movie S2: Overview of sACE-2.99 structure. Coulomb potential density map colored by
subdomain as Figure 1, followed by cartoon-depicted model of sACE-2.99 structure
(colored by sub-domain) fit into its Coulomb potential map (gray), and then cartoon-
depicted model of sACE-2.99 structure (colored by sub-domain).

Movie S3: Overview of sACE-3.05 structure. Coulomb potential density map colored by
subdomain as Figure 1, followed by cartoon-depicted model of sACE-3.05 structure
(colored by sub-domain) fit into its Coulomb potential map (gray), and then cartoon-
depicted model of sACE-3.05 structure (colored by sub-domain).

Movie S4: Overview of sACE-3.15 structure. Coulomb potential density map colored by
subdomain as Figure 1 followed by cartoon-depicted model of sACE-3.15 structure

(colored by sub-domain) fit into its Coulomb potential map (gray), and then cartoon-
depicted model of sACE-3.15 structure (colored by sub-domain).

Movie S5: Non-glycosylated sACE MD simulation. One microsecond excerpt from a
representative sACE all-atom MD simulation without modeled glycans. Protein shown
as cartoon and colored by subdomain. Solvent molecules hidden for clarity.

Movie S6: Glycosylated sACE MD simulation. One microsecond excerpt from a
representative sACE all-atom MD simulation with modeled glycans. Protein shown as
cartoon and colored by subdomain. Glycans shown as gray sticks . Solvent molecules
hidden for clarity.

Movie S7: Non-glycosylated PCA trajectories were generated as linear interpolations
representing the conformational changes described along each of the top 10 principal
components from our analysis (See figure S10). Protein is colored by subdomain.

Movie S8: Glycosylated PCA trajectories were generated as linear interpolations
representing the conformational changes described along each of the top 10 principal
components from our analysis (See figure S11). Protein is colored by subdomain.
Modeled glycans are not shown for clarity, and due to the fact that PCA was performed
using only alpha carbons of protein residues.

Movie S9: Cryosparc 3D variability analysis. Trajectory movies representing structural
variability along the top 5 principal component vectors of variance. Trajectories depict a
volume series of the representative structure generated by averaging particles in every
tenth percentile along the principal component vector.

Movie S10: CryoDRGN heterogeneity analysis. Trajectory movies representing
structural variability along the top 5 principal component vectors of variance.
Trajectories depict a volume series of the representative structure generated by
averaging particles in every tenth percentile along the principal component vector. See
Figure S14 for the path each trajectory follows through PCA and UMAP space.

Movie S11: RECOVAR trajectory. RECOVAR was used to generate a trajectory through
the deconvoluted multidimensional latent space described in Figure 5C, from the sACE-
N/N focus masked analysis. The path was calculated from the latent point representing
the closed-closed state to the latent point representing the open-open state.
Representative structures were generated at six equidistant points along the path and
stitched together as a volume series to generate the trajectory.

Movie S12: 3DFlex trajectory. The most dynamic result from CryoSPARC 3DFlex is
shown. Results were generated using a default tetrahedral mesh and inputting the
modes from 3DVA. Despite this, the motions observed in 3DVA (see movie S9) were not
observed in 3DFlex. The most significant motions observed was a slight opening of
sACE-C (red arrow) and "rocking" of sACE-N/N.