Title

Identification and classification of ion-channels across the tree of life: Insights into understudied

CALHM channels

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Abstract

The ion channel (IC) genes encoded in the human genome play fundamental roles in cellular functions and disease and are one of the largest classes of druggable proteins. However, limited knowledge of the diverse molecular and cellular functions carried out by ICs presents a major bottleneck in developing selective chemical probes for modulating their functions in disease states. The wealth of sequence data available on ICs from diverse organisms provides a valuable source of untapped information for illuminating the unique modes of channel regulation and functional specialization. However, the extensive diversification of IC sequences and the lack of a unified resource present a challenge in effectively using existing data for IC research. Here, we perform integrative mining of available sequence, structure, and functional data on 419 human ICs across disparate sources, including extensive literature mining by leveraging advances in large language models to annotate and curate the full complement of the "channelome". We employ a well-established orthology inference approach to identify and extend the IC orthologs across diverse organisms to above 48,000. We show that the depth of conservation and taxonomic representation of IC sequences can further be translated to functional similarities by clustering them into functionally relevant groups, which can be used for downstream functional prediction on understudied members. We demonstrate this by delineating co-conserved patterns characteristic of the understudied family of the Calcium Homeostasis Modulator (CALHM) family of ICs. Through mutational analysis of co-conserved residues altered in human diseases and electrophysiological studies, we show that evolutionary-constrained residues significantly alter gating functions. Thus, by providing new tools and resources for performing large comparative analyses on ICs, this study addresses the unique needs of the IC community and provides the groundwork for accelerating the functional characterization of dark channels for therapeutic intervention.

41   Keywords: ion channel classification, LLM, RAG, human ion channel orthology, CALHM

42   electrophysiology

Introduction

Ion Channels (ICs) are membrane-bound proteins critical to many physiological processes in the human body, including regulating cell volume, neurotransmitter release, muscle contraction, and glandular secretion [1-4]. Abnormal channel functions have been causally associated with "channelopathies" such as Parkinson's disease, epilepsy, cardiac arrhythmia, cancer, and cystic fibrosis, to name but a few [5-8]. The development of selective chemical probes for channels in disease states is currently hindered by the limited knowledge of the diverse gating mechanisms, ion selectivity, and pathway associations displayed by the various channels encoded in the human genome [8, 9]. While IC sequences from diverse organisms encode critical information regarding underlying functions and effective mining of sequence data can provide important context for predicting and testing understudied IC functions, traditional bioinformatic approaches have had limited success due to the challenges in consistently defining the full IC complement [10, 11], and accurately aligning and mining large sequence datasets [12]. The extensive diversification and widespread abundance of ICs across cell types and their similarities to other transmembrane protein families, such as transporters, have led to an inconsistent definition of the human IC complement. For example, previous literature has reported around 230 human ICs [13]. The Kyoto Encyclopedia of Genes and Genomes (KEGG) database has cataloged around 315 human ICs [14, 15], while Pharos [16] and The Human Genome Organization (HUGO) [17] list around 344 and 330 proteins, respectively. Moreover, these resources classify the human ICs based on their function, which is not known for many. Furthermore, current classification rarely accounts for the pore-forming or auxiliary roles of ICs. Due to these inconsistencies, downstream analyses have been hindered and limited to certain families or groups of closely related channels. Consequently, our current knowledge of IC functions is skewed towards a subset of well-studied "light" channels, while a significant portion of the

channelome remains understudied and is referred to as "dark" channels by the Illuminating

Druggable Genome (IDG) consortium [18, 19].

The current classification of channel genes is largely based on experimentally determined

electrophysiological parameters such as the type of ions they transport (Na, K, Ca, Cl) and the

channel gating mechanism such as gating by voltage potential (voltage-gated) or ligand binding

(ligand-gated), which has been quite helpful in placing ICs in a functional context [20]. However,

selective targeting of channels in diseases and a mechanistic understanding of disease-

associated mutations in channelopathies will require a residue-level understanding of function-

determining features in both well-studied "light" and understudied "dark" channels.

The structure space of the channelome is being rapidly filled in by crystal structures. Cryo-EM

studies and analysis of these structures have shown that despite extensive variation in primary

sequences, several channels adopt common structural folds and mechanisms of action [21]. For

example, several ICs (K, Na, Ca, TRP, and HCN), grouped together as voltage-gated-like

(VGL), are proposed to have evolved from a common ancestor because they share key

sequence and structural features [22, 23]. Thus, quantitative comparisons of these commonly

conserved regions at the sequence and structural level can provide new residue-level insights

into IC function, much like in other large protein families such as kinases and

glycosyltransferases [24-27]. Additionally, a comprehensive evolutionary analysis that includes

orthologous sequences across different organisms from the Tree of Life can add valuable

insights into the evolutionary constraints that govern the conservation of critical functional

residues. Currently, such studies for ICs are limited to a select few closely related organisms

and do not span the full IC complement [28, 29]. Thus, this lack of a unified resource that

systematically identifies and classifies ICs across diverse organisms presents a major challenge

in using evolutionary features for function prediction.

To address these challenges, we first mined literature and sequence data sources to collect

information on ICs stored across various databases and in disparate data sources and formats

to build a well-curated knowledgebase towards defining the human ion "channelome." A

comprehensive list encompassing established and putative IC sequences in the human

proteome is provided as a unified source alongside other contextual data for functional

annotations such as regulatory interactions, ion-selectivity, and pore-lining residues. Based on

this curation, we then perform a comprehensive classification of ICs into well-defined groups

and families and identify the extent of understudied channels in each family, providing a premise

for prioritizing studies in IC families with more understudied members.

We further performed a comprehensive evolutionary analysis to define and collect more than

48,000 well-defined IC orthologs from diverse taxonomic groups spanning metazoans, fungi,

protists, bacteria, and archaea. We demonstrate the application of these sequences in

annotating understudied Calcium Homeostasis Modulator channels (CALHMs) at residue-level

resolution using Bayesian statistical approaches [30] and in predicting and experimentally

validating the structural and functional impact of disease-associated mutations. These studies

support our working premise that integrative mining of available sequence, structure, and

functional data on ICs from diverse organisms (well-studied and understudied) will provide an

important context for defining sequence and structural features associated with understudied

channel functions.

Results

Defining the IC complement of the human genome using informatics approaches

We first sought to collect and curate the full IC complement across the human proteome by

systematically mining all the existing data sources, collating information from literature, and

running various informatics tools (see methods) to identify related sequences based on overall

similarity and pore-defining regions. In brief, we first collected all the sequences listed as ICs in

the comprehensive UniProt database [31], as well as Pharos [16], KEGG [14, 15], Guide to

Pharmacology (GtoP) [32], and HUGO Gene Nomenclature Committee (HGNC) [17] databases

and subjected them to a series of annotations as described in Table 1. The complete annotation

table for all human ICs is available in Supplementary Table 1. Primarily, we mined the literature

for functional and structural annotations on these sequences. Broadly, we have included 6

annotation categories. 1) The identifier labels include the UniProt ID, name, and Target

Development Level (light/dark) designation from Pharos. 2) The classification labels describe

the Group, Class, and Family the IC falls into. This information has been mined from literature

including UniProt, Pharos, GtoP, and HGNC, and the field "Family designation" provides a

consensus family label for the IC. 3) The functional labels have been manually mined from

various sources, relying mostly on previous literature in the "Lit Resource" column. An important

column in this section is the "Unit" column, which describes whether the IC is directly involved in

forming the pore region, where it can be pore-forming (directly involved in forming an ion-

conducting pore), 2-pore (contains 2 tandem pore-conducting regions), or auxiliary (does not

have its own pore domain but interacts with other pore conducting IC subunits as part of a

complex). We also provide the ions, gating mechanism, and UniProt functional description in

this annotation category. To further verify the ion and gating mechanism annotations, we mined

the literature using Large Language Models (LLM) and Retrieval Augmented Generation (RAG)

systems (Methods, Supplementary Figure 1). The RAG system was able to successfully verify

about 40% of the ion and gating mechanism annotations, while for others, it was unable to find

supporting evidence, either because no evidence was available in the existing literature or no

relevant references were found. 4) The structure-related category includes the PDB ID of any

experimentally generated crystal structure or an Alpha-Fold ID for a computationally predicted

structure of the IC. 5) The Complex/Interaction label provides information on the types of

complex the IC is involved in forming, along with the interactors. 6) The last section for TM and

pore domain-related labels has been curated extensively using various sources and predictors.

We start by annotating the TM regions and pores containing functional domains using the

literature and prediction tools, as described in the Methods. TM predictions by TMHMM [33] and

Phobius [34] are provided alongside the TM annotations provided in UniProt. First, we

supplement this information by adding TM information based on the literature review and the

source. Then, we further analyze any ICs with an experimentally generated structure using the

MOLE software [35] to identify and annotate the pore region and pore-lining residues. The

MOLE software starts by defining the membrane region of the protein, followed by the

identification of cavities and computation of the pore boundaries to predict the pore lining

residues. These predictions are used as additional evidence for defining the pore containing

functional domain region and a pore containing IC sequence. To provide functional context to

this classification, sequences, where a pore region was not identified, were classified as

auxiliary following the definition provided by Gurnett et al. 1996 [36]. A snapshot of the list of

annotations we collected through this process is shown in Table 1, using Aquaporin 1 as an

example.

As a final step, we performed sequence similarity searches to identify sequences that share

significant similarity to any of the curated sequences and subjected them to all the annotation

steps to check if any of them qualify as an IC. By combining the results from this annotation

process, we were able to achieve the following: 1) compile a list of human ICs based on the

presence of distinct transmembrane regions, a detectable pore, and evidence of ion

conductivity, 2) distinguish between a pore containing IC vs an auxiliary IC, 3) identify novel

putative IC sequences and 4) curate and classify the full IC complement in the human

proteome. These results are summarized in Figure 1.

419 human ion channel (IC) sequences were curated using this approach, representing an

increase of 75 sequences compared to the previous consensus of 344 ICs in humans [8]. These

sequences were classified into four major groups—Voltage-Gated Ion Channels (VGICs),

Ligand-Gated Ion Channels (LGICs), Chloride Channels, and Others—defined previously [12],

and 55 families. VGICs constitute the largest group, comprising 186 sequences distributed

across 21 families, followed by LGICs with 82 sequences in 10 families. Of the 419 sequences,

62 could not be assigned to the four major groups and were categorized into outlier families.

Among these, 28 are pore-containing ICs, with 19 of the 28 distributed across four families

(CALHM, OTOP, TMC, and TTYH). These families, collectively referred to as "Unclassified"

families, are labeled as such in Figure 1. The remaining 9 pore-containing sequences could not

be assigned to a distinct family and were grouped together under a single "Unclassified" family.

Based on the evaluation of the pore-containing regions, 343 out of 419 ICs were annotated as

pore-containing ICs, while the remaining 76 were classified as auxiliary and fell into 17 different

families. 23 of these auxiliary ICs are soluble with no detectable transmembrane domains. Any

auxiliary ICs that did not belong to a distinct family were collectively classified into an "Auxiliary

unclassified" family.

Next, we sought to use the curated set to define similarities across IC families. Traditional

bioinformatics approaches such as sequence alignment or HMM models could not account for

the vast divergence in primary sequence across IC families and thus not applicable for ICs.

Instead, we relied on representations (embeddings) derived from evolutionary scale protein

language models [37] to capture sequence, structure, and evolutionary information and use

them to generate pairwise sequence alignment. Specifically, for the 343 pore-containing ICs, we

passed their pore-containing functional domains to a protein sequence embedding model called

DEDAL [38] to perform a pairwise sequence alignment. The resulting all-vs-all sequence

similarity scores were used to generate UMAP embedding scores (Supplementary Figure 2). An

average of all IC sequences within a family was calculated to place the IC family bubbles in

Figure 1. Thus, families placed close to each other in Figure 1 indicate their similarities in

sequence embeddings, which capture learned representations of sequence, structure,

evolutionary conservation, and functional properties. Since the auxiliary ICs did not have a

defined domain, they were not used to generate the alignment. Hence, auxiliary IC families are

placed at the bottom of the figure. Most of the VGIC families group together on the top right

except for K_{IR}, K_{2P}, and PLLP, where K_{2P} and PLLP are more centrally dispersed, while K_{IR} falls

closer to LGIC families ASICs, ENAC, and P2X receptor, all of which share similar

transmembrane topology with only two transmembrane helices. The IP3 and Ionotropic

glutamate receptors are also centrally located, while the Cys loop receptor LGIC families group

separately from others towards the bottom right, indicating their shared functional and structural

similarities. Interestingly, Pannexins, VRAC, Connexins, CLCC, and CALHM families group

closer together centrally, indicating shared features.

We also mapped Target Development Level annotations defined by the IDG [18] in Pharos [16]

to highlight the depth of understudied ICs within each group. Briefly, IDG describes Tclin as

targets that have an approved drug; Tchem has known high potency small molecule binding

targets; Tbio has experimentally supported Gene Ontology [39] annotations based on published

literature; while Tdark is manually curated at the primary sequence level in UniProt, but do not

meet any of the Tclin, Tchem or Tbio criteria [16]. There are 19 Tdark, 185 Tbio, 88 Tchem and

127 Tclin ICs spread across all families. The proportion of these levels for each family is shown

as pie charts in Figure 1. CALHM family has the highest proportion of Tdark ICs, with OTOP,

TMC, Connexins, CaCC, CACN, K_{2P}, and the unclassified families accounting for the remaining

Tdark ICs, highlighting the need for more studies on these families. On the other hand, VGICs

K_V, Na_V, Ca_V, RyR, and LGICs $GABA_A$ receptor, nACH receptor, ENAC have the highest

proportion of Tclin ICs, solidifying their status as the well-studied families.

Mining and cataloging IC orthologs across the Tree of Life

Next, we sought to extend this annotation and collect related ICs from other organisms across

different taxonomic lineages from the Tree of Life [31]using the full complement of the annotated

human IC sequences. To this end, we used a graph-based orthology inference approach [40]

starting from the full length and the pore-containing functional domains of the human IC

sequences to define orthologous relationships across more than 1,500 proteomes from the

UniProt Proteomes database [31]. These selected proteomes dataset includes 353 Archaea,

696 Bacteria, and 547 Eukaryota organisms. This method has previously been successfully

employed to define orthology across such a large collection of proteomes for protein kinases

[40], and since it relies on both the full-length sequence and the annotated pore-containing

domains, it can more accurately define true orthologs that are indeed evolutionarily related

across organisms. Since this method relies on the pore containing domain annotations for a

domain-based orthology inference, we only used the 343 conducting IC sequences for this

analysis.

Using this method, we could identify more than 48,000 IC orthologs across diverse organisms.

Figure 2A shows a heatmap depicting the phylogenetic profiling of human IC orthologs along

the horizontal axis across the taxonomic lineages shown by the tree along the vertical axis. The

color intensity indicates the percentage of orthologs detected for all the sequences in the IC

family for a given taxonomic group. The highest number of orthologs were found for Aquaporins,

including bacterial and archaeal orthologs, indicating that this family of ICs is the most

conserved across lineages. Along with Aquaporins, the two paralogs of Golgi pH regulator ICs

(GPHRs), both indicated as Tbio channels, were found to have the largest number of orthologs

across metazoans. Along with GPHRs, several other ICs, notably members of the K_V family and

Gamma-Aminobutyric acid type A (GABA$_A$) receptors, display some of the most widespread

orthologs across organisms. Some dark channels, such as members of the TMC (TMC7 and

TMC3) and CALHM (CALHM5 and CALHM3) families, are well conserved across metazoans

and vertebrates, respectively, yet very little is known about their functions. On the other hand,

some Tdark channels, such as Potassium channel K member 7 (KCNK7) and Transmembrane

channel-like protein 4 (TMC4) have a lineage-specific set of orthologs that extend only up to

mammals. A full list of all the orthologs detected through this analysis is provided in

Supplementary Table 2.

To use this evolutionary conservation for functional inference of dark ICs, we first used

hierarchical clustering to group ICs with similar orthology profiles into related clusters, which led

to the definition of 9 clusters, as shown in Figure 2B. Since the orthology profiles in prokaryotic

lineages were very sparse, only the eukaryotic orthology profiles were retained for this analysis.

Within eukaryotes, each cluster has a well-defined signature of orthology conservation. For

example, ICs in cluster 2 have most of their orthologs only in mammals, whereas clusters 4, 5,

and 6 have orthologs spanning other vertebrates. Similarly, clusters 7 and 8 have orthologs

extending to other metazoans, while seven ICs in cluster 9, including the 2 Golgi pH regulators

A and B and the sodium leak channel NALCN, have detectable orthologs in Fungi, Plants,

Protists, and other eukaryotic lineages. To translate these patterns of orthology profiles to

function, we performed a functional GO term enrichment analysis for the human ICs in each

cluster (Figure 2C). As expected, the most significant GO terms were related to ion channel

function (colored green in Supplementary Table 3). However, beyond channel function, the

enriched functional GO terms obtained for each cluster correlate well with the physiological

functions present in the orthologous organismal groups. For example, clusters 4, 5, and 6 are

conserved in vertebrates and have six dark IC sequences, including unclassified ICs such as

CALHM 3 and 5, and connexins GJA10, GJD4, and GJE1, which had functional enrichment for

regulation of heart contraction and heart rate, traits that could be closely related to a closed

blood circulatory system with an endothelium, present in vertebrates [41]. On the other hand,

cluster 2 with orthologs in mammals was conserved in mammalian-specific reproductive

functions such as sperm capacitation.

Using the orthology information to identify evolutionary constraints in CALHMs

Once the orthologs have been identified across different organisms, they can be leveraged to

find evolutionarily conserved signals that could point to functional similarities within related

groups of sequences. To this end, we classified subsets of orthologous IC sequences into

evolutionarily related clusters using a Bayesian Partitioning with Pattern Selection (BPPS)

algorithm, which classifies sequences based on patterns of amino acid conservation and

variation in a large multiple sequence alignment (see methods) [30]. For this analysis, we

focused on the orthologs of CALHM family of IC sequences. The CALHM proteins constitute a

family of large pore channels, forming oligomeric assemblies of different sizes [42]. It has 6

member sequences in humans (CALHM1-6), 3 labeled as dark channels, constituting one of the

families with the highest prevalence of dark ICs. It has been well established that CALHM1 is

activated by removing extracellular Ca and membrane depolarization and that the heteromeric

CALHM1 and CALHM3 channels are implicated in the ATP release during taste perception [43].

In contrast, the activation stimuli and physiological roles of other family members remain largely

unexplored. All CALHM family members share a common arrangement in the transmembrane

domain, with each subunit consisting of four transmembrane segments (S1–4), forming a large

cylindrical pore lined by the first transmembrane segment S1 along with a short N-terminal helix

preceding S1. While the exact gating mechanism of the CALHM channels is still unclear, it is

hypothesized that the conformational state of S1 plays a crucial role in determining the pore

size.

Figure 3A shows a phylogenetic tree depicting the evolutionary relationship across the 6

members of the CALHM family. CALHM1 and 3 form a distinct clade from CALHM 2, 4, 5 and 6.

We first analyzed 5805 CALHM homologs to identify pattern positions conserved across all

these sequences with the hypothesis that such conserved positions could point to shared

functional features across all CALHMs. The Bayesian analysis identified 13 aligned positions

conserved across all 6 CALHM homologs, which we will refer to as CALHM shared patterns

(Figure 3B,C). Most of these conserved positions were hydrophobic amino acid residues, and 5

conserved Cystine residues, 4 of which are involved in forming inter-molecular disulfide bridges

(C46=C130, C48=C162). Residue position numbers reflect the numbering based on human

CALHM2 (PDB id: 6uiv). Interestingly, most other conserved positions (F44, Y56, I61, P64,

W117) are located around a functionally important linker region connecting S1 and S2 (S1-S2

linker). Specifically, this linker region could regulate the dynamic conformational changes of S1,

where the S1 could adopt either a vertical conformation relative to the membrane plane,

resulting in an enlarged pore size, or a lifted conformation, leading to a reduced pore size

(Figure 3D) [44]. Therefore, mutations in this linker are expected to affect channel functions, and

the conserved pattern positions identified in this analysis are hypothesized to play a role in the

gating mechanism of CALHM. The conservation of these residues across orthologs of all 6

CALHM sequences further suggests that all CALHM paralogs could share this gating

mechanism.

To determine the functional importance of the CALHM shared patterns, we sought to perform a

series of mutational experiments to determine the functional implications of perturbations at

these positions. To achieve this, we methodically determined target mutations for each position.

We first scanned the Genome Aggregation Database (gnomAD) [45] to check for any prevalent

variations at these conserved positions within the sampled population to use as our mutational

targets. We found several disease variants at these positions that are listed in Figure 3E that

were used to prioritize target mutations. For positions where a variation was not found, we

tested their significance by performing alanine mutations, causing a deletion of the side chain at

the β-carbon.

Experimental studies to probe the functional implications of the evolutionary constraints

To assess the functional roles of the predicted conserved residues in CALHM channels, we

performed several informed mutational analyses of these residues in representative human

CALHMs and subjected them to patch clamp electrophysiology studies. We selected the well-

studied human CALHM1 and the understudied human CALHM6 for our analyses. The

expression levels of wild-type proteins and mutants are comparable, as judged by western blot

and in-gel fluorescence images (Supplementary Figure 3). We established electrophysiological

measurements for wild-type CALHM1 and CALHM6 following protocols described in the

methods. Our results were consistent with previous findings [46] (Figure 4). In addition to room

temperature, which is commonly used for patch-clamp studies of CALHM channels, we also

conducted measurements at physiological temperatures (37°C). We observed that CAHLM1

activity was significantly higher at physiological temperature than at room temperature, while the

CALHM6 currents showed only a small increase (Figure 4). The robust currents at 37°C,

particularly those of CALHM1, provide a solid basis to interpret the phenotypes of the mutants

tested, as detailed below.

Two predicted conserved residues are in the S1–S2 linker (F40 and Y52 in CALHM1; F39 and

Y51 in CALHM6). As expected, mutations in these residues either abolished or markedly

reduced channel activity in both CALHM1 and CALHM6, presumably by impeding the

conformational dynamics of S1 required for channel gating (Figure 5). Interestingly, mutations in

other conserved residues, not within the S1–S2 linker but near it (W114, L57, and P60 in

CALHM1; W113 and A127 in CALHM6), also resulted in abolished or markedly reduced channel

activity (Figure 5). Among these residues, W114 (of CALHM1) and W113 (of CALHM6) have

direct interaction with S1, thus changing its bulky hydrophobic side chain into a small side chain

(cysteine or alanine) or positively charged residue (arginine) may significantly alter the

conformational dynamics of S1 and therefore impair channel gating. Our experimental data

indicate that the conserved pattern residues at the S1–S2 linker and its immediate surroundings

are important for the gating mechanism across CALHMs.

Discussion

Here, for the first time, we have computationally defined the full IC complement of the human

genome – the "channelome". We provide this comprehensive list and a rich annotation of

functional, structural, and sequence features, mapping them to 4 widely accepted IC groups and

further classifying them into 55 families. We also highlight unclassified outlier groups that

contain most of the understudied "dark" and potentially novel unclassified IC sequences. As part

of our curation, we also provide annotation for the pore-containing domain that is based on

multiple sources, including an extensive literature review, further strengthened by the

application of literature mining using LLMs and structural analysis tools. We use the pore-

containing domain to distinguish pore-containing from auxiliary ICs, providing additional

functional context. Our annotations can serve as a reference for comparing ICs within or across

families, designing experiments for functional studies, and identifying potential targets to

illuminate the functional relevance of understudied ICs further.

The application of LLMs, specifically RAG systems, in our research demonstrates significant

potential in automating literature mining and correcting certain inaccuracies. However, the

system currently faces limitations related to diverse data presentation formats, complex channel

subunit relationships, and variant gene nomenclature, which hinder its effectiveness. The high

frequency of responses where evidence was not found underscores the need for enhanced

entity recognition capabilities, particularly for accurately identifying gene and protein names

across different nomenclatures. Additionally, the RAG system's ability to critically engage with

data by identifying and amending presumed inaccuracies can be widely used to verify and

correct annotation entries, though this also raises concerns about potential overcorrections or

misinterpretations without adequate validation. By more data cleaning and standardizations,

LLMs and RAG systems can become more reliable and integral tools for curating and

expanding bioinformatics databases, ultimately enhancing the accuracy and

comprehensiveness of our IC knowledgebase.

To the best of our knowledge, identifying true orthologs across the Tree of Life performed here

is the first large-scale effort to consolidate ion channel orthologs across diverse organisms and

provides a comprehensive view of the evolutionary conservation of individual ICs. Based on this

analysis, we could pinpoint the Aquaporins family of ICs as the most ancient family of ICs with

orthologous sequences present back to bacterial and archeal species. More importantly, the

clustering of human ICs based on their depth of conservation placed them into functionally

meaningful clusters with many enriched functions exclusive to their orthologous taxonomic

group. The placement of understudied ICs in such groups allows for meaningful functional

predictions, which can be further elucidated using experimental approaches.

The orthologous sequences identified and presented here can serve as a valuable resource for

performing evolutionary and functional analysis using statistical and machine learning

approaches. We demonstrate one such use by performing a Bayesian pattern-based

classification on the CALHM subset of ICs to identify amino acid positions significantly

conserved across all CALHM sequences. These features reside on a functionally important S1-

S2 linker region that potentially governs their gating mechanism. Because these features were

conserved and shared across all CALHM sequences, we also hypothesize that all CALHM

sequences share this gating mechanism. By performing targeted mutations on these conserved

residues, we show that a mutation in any of these evolutionarily conserved residues results in a

dramatic loss of gating function, thus highlighting the functional importance of the identified

pattern positions. This was shown not just on the well-studied CALHM1, but also on the

understudied human CALHM6, thus providing valuable clues into the shared function of light

and dark CALHM sequence sets.

Among the conserved residue set, the positioning of P60 on the S2 helix of CALHM1 is

intriguing. In the cryo-EM structure, P60 resides approximately midway along S2, inducing a

distortion or bending of the helix due to the rigid ring structure of proline. This configuration

allows extensive contact between the extracellular portion of S2 and the S1–S2 linker, while the

intracellular portion of S2 interacts extensively with S1. Furthermore, we hypothesize that this

bent conformation of S2 contributes to the flexibility of the S1–S2 linker. Because if S2 were

straightened, it would consequently straighten the S1–S2 linker, potentially reducing its

flexibility. Replacing the proline with glycine eliminates the structural constraint imposed by

proline's cyclic side chain, allowing S2 to adopt a more regular, uninterrupted helical structure.

Consequently, a straightened S2 may lead to reduced flexibility of the S1–S2 linker, ultimately

impacting channel gating. L57 is positioned near P60 on the protrusion of the bend of S2, with

its sidechain facing residues on the adjacent S3. We hypothesize that the larger sidechain of the

L57R mutant would force S2 to straighten due to steric collision between the bulky arginine

sidechain and residues on S2, thereby reducing the flexibility of the S1-S2 linker. A127 resides

in a short alpha helix within the extracellular domain. While A127 does not directly interact with

the S1–S2 linker, it is adjacent to the highly conserved disulfide bond (between C41 and C126),

linking the helix containing A127 to the S1–S2 linker. Thus, it is conceivable that the A127G

mutant may indirectly affect the conformation of the S1–S2 linker through this disulfide bond.

Since all mutants showed either reduced or completely abolished activity, we included a positive

control in the patch clamp experiments: I109W on CALHM1, which has been previously

documented to increase channel activity and were able to reproduce this phenotype (Figure 5).

We also examined the corresponding mutant in CALHM6, L108W (I109 in CALHM1 aligns with

L108 in CALHM6), to further explore channel-specific differences. Interestingly, L108W

decreased CALHM6 channel activity (Figure 5), possibly reflecting inherent differences between

the two channels.

In addition to the CALHM shared patterns, we also present a second subset of conserved

positions shared only within CALHM 2,4,5 and 6, paralogs that are evolutionarily closer

compared to CALHM1 and 3 (Supplementary Figure 4). These subsets of conserved residues

fall in the intracellular helical segment that is involved in oligomerization and formation of the

pore complex governing CALHM function. Previous studies have indicated that CALHM2 adopts

a unique undecameric oligomer different from the CALHM1 octamer [47]. Thus, with the

presence of shared patterns within the subset of CALHM2 and the understudied CALHM4,5 and

6 in this intracellular helical region, we hypothesize that these conserved positions help maintain

a similar mode of oligomerization between these evolutionarily related subsets of CALHMs.

Because many disease mutations map to these conserved positions, we believe the datasets

and approaches offer a powerful avenue for elucidating the functional and clinical relevance of

the understudied ICs.

Methods

Identification and annotation of human ICs

The annotation of human ICs was performed using a semi-automated pipeline. First, all the

protein sequences that were labeled as ICs were collected from the UniProt [31], KEGG [14,

15], Pharos [16], GtoP [20], and HGNC [17] databases. The annotations described in Table 1

were then collected from literature sources manually and compiled together. The UniProt

specific labels and the complex formation information were extracted from the UniProt

database. The sequences were run through TMHMM [33] and Phobius [34] to predict

transmembrane and helical regions. The predictions were cross-referenced and confirmed

against the CDD [48], Pfam [49], PrositePattern , PrositeProfiles [50] and SMART [51]

databases, where a reference was available. Finally, the prediction of the pore region and pore-

lining residues was supplemented using the MOLE software [35].

RAG system to verify ion specificity and gating mechanism annotations

To enhance the curation and annotation of human ion channels, we developed a Retrieval-

Augmented Generation (RAG) system that leverages large language models (LLMs) to verify

existing annotations. This method combines vector databases for efficient information retrieval

with the natural language understanding capabilities of LLMs, allowing for systematic verification

of IC annotations. The system can identify discrepancies or confirm existing information based

on the latest available literature. The modular design of this approach facilitates future

improvements and adaptations to other annotation verification tasks in bioinformatics.

We began the data collection by downloading relevant PubMed articles to each annotation,

storing the PDF files locally, and converting them to text. After filtering out irrelevant content, the

extracted text was split into manageable chunks. To create the vector database, we utilized

OpenAI's text-embedding-ada-002 model to generate embeddings for each text chunk. Each

processed and embedded chunk was then added to the vector database collection, with

metadata including the source PubMed ID attached to each vector.

We designed the query system to construct two queries for each IC: one for verifying ion

selectivity and another for confirming the gating mechanism. These questions were dynamically

generated based on the manually curated annotation table. The RAG system architecture

performed similarity searches in the vector database using the query text, conducting filtered

(using specific PubMed IDs) and unfiltered searches to ensure comprehensive retrieval. The top

three most relevant chunks from the search results were combined to form the context for the

LLM. A custom prompt template was designed to guide the LLM in providing evidence-based

answers, including instructions for JSON-formatted output and emphasizing the importance of

evidence-based responses.

For LLM integration, we passed queries to OpenAI's GPT-4 model. The LLM's responses were

parsed and structured into a JSON format, including an answer (*Found* or *Not Found*), a

confidence value between 0 and 1, and supporting evidence from the retrieved context. This

structured output allows easy integration with existing annotation data, enabling systematic

comparison and updating of ion channel information. Where the LLM suggested alternative ion

selectivity or gating mechanism, such instances were manually verified and corrected. Finally,

the ion selectivity and gating mechanism annotations not supported by LLM were labeled in our

annotation table (Supplementary Table 1).

Defining a pore containing functional domain

We aimed to define the pore-containing functional domain so that it spanned the pore region

and included all the TM domains present in an IC. This ensured that our domain definition

included the pore region and any other functionally important TMs, while excluding any

accessory domains on the flanking sequences. To ensure we did not miss any TM regions, four

different sources of annotation information were combined to define the pore containing

functional domain for the 343 ICs. 1) The TM predictions from TMHMM and Phobius was first

used to identify the TM regions. 2) The UniProt-based TM annotations were matched with the

predictions to verify the TM regions further. 3) The literature-based TM annotations were then

used to verify the TM positions and organization, where available manually. 4) Where

experimentally resolved crystal structure coordinates were available, the MOLE software was

used to identify the pore lining residues.

Pairwise sequence alignment using sequence embeddings

The pore containing functional domains for the 343 pore containing ICs were passed to DEDAL

[38] which was run using standard parameters and resulted in homology logit scores based on

an all-vs-all pairwise sequence alignment using their sequence embeddings. These scores were

used to generate a sequence similarity matrix passed to the umap function of the umap_learn

package v0.5.7 [52] in python 3.9 to generate 2D UMAP embeddings. This was used to

generate a 2D scatterplot that defined the placement of individual ICs shown in Supplementary

Figure 2. Finally, an average position of ICs within a family was used to define the placement of

that family in Figure 1.

Orthology detection and analysis

The KinOrthto pipeline [40] was used for defining the orthologs and co-orthologs. In brief,

KinOrtho uses a graph-based orthology inference approach using both the full-length and the

pore domain regions for inferring orthologous relationships primarily relying on sequence

similarity within these regions. The pipeline began with a homology search using both full-length

and pore-containing domain sequences of human ICs against the proteomes database. Since a

pore domain needs to be defined for this analysis, we only used the 343 pore containing IC

sequences to perform orthology detection. They were used as query for running KinOrtho

against a reference database of curated proteomes from the UniProt Proteomes Release

2022_05 that consisted of 343 Archaea, 696 Bacteria and 548 Eukaryota proteomes. After

running reciprocal BLAST searches between the query and the hits, orthology detection was

performed, followed by cluster analysis using OrthtoMCL and filtering of relationships to keep

only relationships within the same cluster. Finally, the results obtained from both the full-length

and pore domains were retained as true orthologous relationships to remove extraneous hits.

48694 unique orthologous relationships were defined and then used to create a

presence/absence matrix of orthologs, with rows representing each human IC and columns

representing the different organisms. Firstly, the orthologs were grouped by IC families and

defined taxonomic lineages. A percentage value representing the proportion of detected

orthologs was calculated for each IC family and lineage. This was calculated using the following:

(total number of orthologs found for all ICs in a family)/(total number of organisms queried in the

taxonomic lineage * number of sequences in the family).

Only eukaryotic orthologs were retained for the orthology profiling clustering and enrichment

analysis. Further, only organisms with an ortholog for at least one human IC were included. This

was used to perform hierarchical clustering with the Ward method of clustering and Euclidean

distance metric using the scipy package [53] in Python. To group ICs with similar

presence/absence patterns, we broadly defined 9 clusters based on the dendrogram. Human

ICs falling in these 9 clusters were then subjected to a functional enrichment analysis using the

Gene Ontology resource GO enrichment tool. GO terms with a corrected FDR <0.01 were

retained as significantly enriched terms.

CALHM evolutionary analysis

BPPS was used to perform a pattern-based classification of the CALHM homologs. First,

orthologs for all 6 human CALHMs were collected. This ortholog dataset was supplemented with

more hits from the UniProt database using MAPGAPS, a multiply-aligned profile for global

alignment of protein sequences [12]. Along with finding the best hits for the human CALHMs,

MAPGAPS also aligns those hits to the template profile alignment to generate a large multiple-

sequence alignment of the resulting 5805 CALHM. This large alignment was then subjected to

BPPS, which performs a hierarchical classification of the sequence sets based on conserved

pattern positions shared by subsets of sequences using a Bayesian statistical procedure [30].

This generates a hierarchical cluster where sequences within each cluster are defined by

distinct conserved patterns.

CALHM construct design, transfection, and western blot analysis

Full-length human CALHM1 and CALHM6 in the pEGC Bacmam vector was used. The

translated product contains the human CALHM1 or CALHM6 protein, a thrombin digestion site

(LVPRGS), an enhanced GFP protein, and an 8x His tag. Primers for site-directed mutagenesis

were designed using Snapgene and synthesized by Eurofins Genomics. The QuikChange

mutagenesis protocol was used to generate all the mutants of the study. Sanger sequencing

was performed to identify positive clones.

Adherent HEK293T (ECACC, Catalogue Number: 96121229) cells were grown in DMEM media

supplemented with 10% fetal bovine serum. Transient transfection was conducted using

lipofectamine-2000 by following the manufacturer's protocol. Specifically, the cells were cultured

in 60 mm Petri dishes until 80% confluency. Transfection solution was made by mixing 500 ng

of plasmid DNA, 4 uL of lipofectamine-2000 reagent, and 100 uL Opti-MEM media. After 20 min

incubation at room temperature, the DNA-lipid complexes were added to the cell culture and

incubated at 37°C. The next day, 10 mM sodium butyrate was added to the cells to boost

protein expression. The cell culture was then grown at 30°C for another day before harvesting.

The cell pellet was flash-frozen with liquid nitrogen and stored at -80°C. Each CALHM1 and

CALHM6 mutant were transfected in triplicate as biological replicates.

For the western blot experiment, the cell pellet was lysed in TBS buffer (20 mM Tris pH 8.0, 150

mM NaCl) with 10% lauryl maltose neopentyl glycol and cholesterol hemisuccinate detergent on

ice. The lysate was then solubilized at 4°C for 1 hour before being clarified by centrifugation at

13,000 rpm for 5 min. The soluble portions were mixed with 4X SDS loading buffer

supplemented with 5% 2-mercaptoethanol. The samples were resolved in precast gradient SDS

gel (4-20%). A Chemidoc instrument was used to directly image the in-gel fluorescence signal

after electrophoresis to obtain the GFP signal from the CALHM1 and CALHM6 proteins.

Subsequently, the protein in the gel was transferred to the nitrocellulose membrane using the

semi-dry transfer buffer (48 mM Tris base, 39 mM glycine, 20% methanol). The membrane was

blocked in the TBST buffer (20 mM Tris pH 8.0, 150 mM NaCl, 0.1% Tween 80) with 4% non-fat

milk for 1 hour at room temperature. Afterward the 1-step beta-actin antibody (1:2000 dilution)

was incubated with the membrane for 1 hour at 4°. After incubation, the membrane was washed

with TBST buffer 4 times, 15 mins each. The western blot signal was detected using ECL Pierce

substrate and imaged using a Chemidoc instrument. A brightfield image was overlaid with the

illuminance signal to visualize the position of protein markers in the membrane. The anti-beta-

actin antibody was obtained from Proteintech (Catalogue number: HRP-60008). Each mutant

was run in a western blot in triplicate from the transfected cell pellets as biological replicates,

with normalization by beta-actin signal as a housekeeping gene.

For quantification of the western blot data, ImageJ was used to make measurements of the

signal from the GFP fluorescence for CALHM1 and CALHM6 and to normalize the data with the

chemiluminescence signal from the beta-actin blot. Measurements were taken through ImageJ

of each lane of protein signal, then divided by the beta-actin signal as a loading control to

normalize the data. The signal for each mutation was averaged, and the standard estimated

mean was calculated before comparing it to the signal from the wild-type CALHM1 or CALHM6

protein. All mutants for each protein were compared in Microsoft Excel using a bar graph with

the standard estimated mean as the error bar values.

Electrophysiology

TsA201 cells expressing plasmids encoding N-terminal GFP tagged human CALHM1 or

CALHM6 were used. After 1-day post-transfection with plasmid DNA (100 ng/mL) and

Lipofectamine 2000 (Invitrogen, 11668019), the cells were trypsinized and replated onto poly-L-

lysine-coated (Sigma P4707) glass coverslips After cell attachment, the coverslip was

transferred to a recording chamber. Whole-cell patch-clamp recordings were performed at room

temperature (21-23°C) or body temperature (36-38°C). Signals were amplified using a

Multiclamp 700B amplifier and digitized using a Digidata 1550B A/D converter (Molecular

Devices, Sunnyvale CA). The whole-cell current was measured on the cells with an access

resistance of less than 10 MΩ after the whole-cell configuration was obtained. The amplifier

circuitry compensated the whole-cell capacitance. The two-step pulse from 120 to -80 mV for 50

ms was continuously applied to the cell membrane every 5 sec to monitor the activation of the

CALHM current. The step pulse from -100 to 140 mV for 200 msec with a holding potential of 0

mV was applied to plot the current-voltage relationship. Electrical signals were digitized at 10

kHz and filtered at 2 kHz. Recordings were analyzed using Clampfit 11.3 (Axon Instruments

Inc), GraphPad Prism 10 (La Jolla, CA), and OriginPro 2024 (OriginLab, Northampton, MA). The

standard bath solution contains (in mM): 150 NaCl, 5 KCl, 1 $MgCl_2$, 2 $CaCl_2$, 12 Mannitol, 10

HEPES, pH=7.4 with NaOH. For a whole-cell recording, the extracellular solution contains (in

mM): 150 NaCl, 10 HEPES, 1 $MgCl_2$, 5 $CaCl_2$. 5 mM $CaCl_2$ was replaced with 5 mM EGTA. The

intracellular solution contains (in mM): 150 NaCl, 10 HEPES, 1 $MgCl_2$, 5 EGTA.

All data are expressed as mean±SE. Multiple comparisons were performed by one-way or two-

way ANOVA with Bonferroni's post hoc test. n indicates the number of cells. Significance was

defined as: *$P<0.05$, **$P<0.01$, ***$P<0.001$. The absence of an asterisk indicates non-

significance.

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641   Conceptualization: RT, NK, WL. Methodology: RT, NK, SJP, SK, NG, SS, WL. Formal analysis

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645   WL.

Competing Interests Statement

648   The authors declare no competing interests.

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 Figures and Tables

Identifier labels		Structure related labels			# of TMs	6
Uniprot	P29972	PDB ID	8CT2		# of Transmembranes (TMs) + Intramembranes (IMs)	10
Name	Aquaporin-1	Alphafold ID			TM Start (UniProt)	8
Symbol	AQP1 (CHIP28)	Complex/Interaction related labels			TM End (UniProt)	228
Target Development level (Pharos)	Tbio	Auxiliary	No		TMsList	T:8-36,T:49-66,I:71-76,I:77-84,T:95-
Length	269	Characterized domains	MIP		TM Lit resource	https://scholar.google.co
Classification labels		Auxiliary domain	no		Lit based TMstart_1	8
Family designation	Aquaporin	Auxiliary protein			Lit based TMend _1	36
Group	Other	Notable interactors	EPHB2		Lit based TMstart_2	49
Class		TM and pore domain related labels			Lit based TMend_2	66
Family	Aquaporin	Pore Domain start	8		Lit based TMstart_3	95
Sub Family		Pore Domain end	228		Lit based TMend_3	115
Functional labels		Domain length	220		Lit based TMstart_4	137
Unit	pore-containing	Does it pass through membrane at least once	Yes		Lit based TMend_4	155
Ion	water	# of TM domains (predicted by TMHMM)	6		Lit based TMstart_5	167
Gate Mechanism	ligand-gated (cGMP)	# of TM domain (predicted by Phobius)	6		Lit based TMend_5	183
Lit Resource	PMID: 26365508, PMID: 16962972	TM organization	1|2|3|4|5|6		Lit based TMstart_6	208
Uniprot Function	Form water-specific channel / plasma membranes of red cells and kidney proximal tubules	MOLE pore residue first	35		Lit based TMend_6	228
		MOLE pore residue last	185			

828 Table 1: List of features annotated for the collected IC sequences. The labels are colored by

829 their annotation category. Labels for Aquaporin 1 are shown as examples of each annotation

830 label.

Figure 1: Distribution of ICs across different families. Each circle represents an IC family, with the symbol at the center indicating its Group. The size of the circles is proportional to the number of sequences in that family, and the colored pies indicate the proportion of their studied status as designated by IDG. Families are placed based on their distribution of UMAP embeddings generated using protein embedding-based pairwise sequence alignments (Supplementary Figure 2).

Figure 2: Orthology profiling of human ICs. (A) Heatmap showing the percent of orthologs

detected for each IC family within a given taxonomic lineage. The taxonomic groups are shown

in the vertical axis with a tree on the left. Darker color represents a higher percentage of

orthologs detected. Percentages were calculated as (total number of orthologs found for all ICs

in a family)/(total number of organisms queried in the taxonomic lineage * number of sequences

in the family) (B) Clustergram depicting the presence/absence of orthologous sequences of ICs

across eukaryotic taxonomic lineages. ICs are clustered along the horizontal axis into 9 distinct

clusters. Taxonomic groups are shown on the vertical axis. Each square in the heatmap is

colored based on the orthology relationship found for a specific IC in a specific organism (black:

one-to-one ortholog present, red: co-ortholog detected, brown: no orthology detected). (C)

Results from the enrichment analysis performed on human ICs of each cluster. The x-axis

shows the number of ICs in the cluster enriched for the GO term shown on the y-axis. The bars

are colored based on their FDR values for the enriched term. For a full list of enriched terms,

please refer to Supplementary Table 3.

Figure 3: Evolutionary analysis of CALHM reveals conserved pattern positions. A) A

phylogenetic tree depicting the evolutionary relationships across all 6 CALHM paralogs with

representative sequences from humans, mouse and rat. B) Schematic representing the location

of transmembrane regions and identified conserved pattern positions in a representative human

CALHM2 sequence. C) The conserved pattern positions conserved across all 6 CALHM

paralogs are shown as a weblogo with the red bars indicating the significance (longer bars

indicate higher significance) and are mapped into a representative structure of human

CALHM2. D) Cartoon representation of CALHM2 structure (PDB ID: 6uiv and 6uiw) in open and

closed conformation, respectively. E) List of disease variants and mutations performed in the

conserved pattern positions for functional studies.

Figure 4: Electrophysiological studies of human CALHM1 and CALHM6. A,C,E. Representative

current traces in the presence of 5 mM Ca2+ in the bath solution at 22°C (left), 0 mM Ca^{2+} at

22°C (middle), and 0 mM Ca^{2+} at 37°C (right) from non-transfected tsA201 cells (A), or tsA201

cells overexpressing wild-type CALHM1 (C), and wild-type CALHM6 (E), recorded in the whole-

cell patch-clamp configuration. Voltage clamps were applied from −100 mV to 140 mV (200 ms

each step) with a final tail pulse at −100 mV (200 ms), with a holding potential of 0 mV. B,D,F.

The mean current amplitudes at the end of each pulse step of the experiments in (A,C,E) were

plotted as a function of clamp voltage. Error bars represent SEM (CALHM1: n=5, CALHM6:

n=3).

Figure 5: Current amplitudes of wild-type CALHM1, wild-type CALHM6, and mutants at conserved pattern positions in human CALHM1 and CALHM6. Currents were recorded at 120 mV in the presence of 5 mM Ca^{2+} in the bath solution at 22°C, 0 mM Ca^{2+} at 22°C, and 0 mM Ca^{2+} at 37°C. Bars represent the mean. One-way ANOVA with Bonferroni's post hoc test was used for statistical analysis (*p<0.05. **p<0.01, ***p<0.001).