AACDB: Antigen-Antibody Complex Database — a Comprehensive Database Unlocking Insights into Interaction Interface

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Abstract

Antibodies play a crucial role in the vertebrate immune system and have diverse biomedical applications. Understanding the intricate interactions between antibodies and antigens is an important step in drug development. However, the complex and vast nature of the data presents significant challenges in accurately identifying and comprehending these interactions. To address these challenges and enhance our understanding of the antibody-antigen interface, the Antigen-Antibody Complex Database (AACDB) has been established. The current version provides a comprehensive collection of 7,498 manually processed antigen-antibody complexes, ensuring accuracy and detail. This database not only provides rich metadata but also corrects annotation errors observed in the PDB database. Furthermore, it integrates data on antibody developability and antigen-drug target relationships, making it valuable for assisting new antibody therapies development. Notably, the database includes comprehensive paratope and epitope annotation information, thereby serving as a valuable benchmark for immunoinformatics research. The AACDB interface is designed to be user-friendly, providing researchers with powerful search and visualization tools that enable effortless querying, manipulation, and visualization of complex data. Researchers can access AACDB completely online at http://i.uestc.edu.cn/AACDB. Regular updates are promised to ensure the timely provision of scientific and valuable information.

Key words: Antibody; Antigen; Interaction Interface; Benchmark; Database

1. Introduction

Antibodies are not only an essential part of the vertebrate immune system, but also has wide application in biomedical field [1, 2]. In clinical practice, the emergence of monoclonal antibodies (mAbs) has revolutionized the treatment for various cancers, autoimmune diseases, and many other diseases. Due to their large binding interfaces, high affinity and specificity, antibody drugs have successfully targeted numerous proteins that were previously impervious to small molecule therapies. The expansion of antibody drug formats, such as antibody-drug conjugates (ADCs), domain antibodies, bispecific antibodies and antibody fusion proteins, has broadened their indications to a wider range of diseases including blood disorders, infections, neurological conditions, ocular diseases, and metabolic disorders [3-5]. The process of antibody drug development can be briefly divided into three stages: preclinical research, clinical trial and post-marketing surveillance study [6]. The preclinical research of antibodies mainly refers to a series of in vitro experiments and animal model studies conducted before the application of

antibodies in clinical trials. In this stage, an important task is to determine the binding interfaces
and interaction sites between antibodies and their drug targets. Understanding how antibodies
specifically interact with their targets or antigens can enable help optimize antibody drug leads
further, as well as provide insights into mechanisms of interaction [7].

The field of machine learning and deep learning-based methods for antigen and antibody
binding interfaces prediction have gained significant attention in recent years [8-11]. The
advancement of this field heavily relies on a reliable foundation of data. While the Protein Data
Bank (PDB) [12, 13] is a well-known repository for protein complexes, the identification of
antigen-antibody complexes from the vast pool of general protein structures remains challenging.
In 2018, AbDb [14] was introduced as a specialized database for antigen-antibody complexes.
However, there have been no further updates to AbDb since then. It is crucial to emphasize that
our statistical analysis reveals a significant increase in the number of antigen-antibody complex
entries within the PDB database during the period from 2021 to 2023. These entries accounted for
approximately 45% of the total antigen-antibody complex entries. The AbDb database currently
lacks proper organization and management of this valuable data. In comparison, the series of
SAbDab databases [15-17] indeed provide a comprehensive data source of antibody structure and
achieve in time updating. However, this rapid updating process may inadvertently overlook a
significant amount of information that requires thorough verification, thereby increasing the
likelihood of incorporating erroneous annotations directly from the PDB database.

Furthermore, another issue in the field of interface prediction is the definition of interacting
amino acids. Some datasets are based on differences in the solvent accessible surface area
(ΔSASA) for each residue upon binding [9, 18]; Others are based on distances between atoms [8].
Different studies employ different datasets and definitions, resulting in a lack of comparability.
There is an urgent need for a standard benchmark dataset to enable objective comparisons of these
methods. None of the databases mentioned above provide details of interaction residues.
Motivated by these issues and challenges, we have developed the Antigen-Antibody Complex
Database (AACDB). It has a user-friendly interface for convenient querying, manipulation,
browsing, and visualization of comprehensive information about antibody-antigen complexes.
Notably, AACDB also provides detailed information on interacting residues using two methods
(ΔSASA and atom distance), facilitating the benchmarking studies of the predictive methods for
paratopes and epitopes.
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2. Materials and Methods

78   We search the PDB database using the keywords "antibody", "antigen", "immunoglobulin"
79 and "complex". Over 32000 experimental structures were extracted (up to December 2023). Each
80 entry was processed following the steps in Figure 1.

2.1 Entry screening

We aimed to retrieve all antibody-related complexes from the PDB using the chosen search
approach. However, we encountered a notable number of false hits in the results, prompting us to
implement a stringent structural filtering process. Antibody complexes were defined as structures
containing at least one antibody molecule and another large protein (with a length greater than 50
amino acids). Structures that lacked antibodies or consisted solely of one type of antibody were
excluded from further analysis. It is crucial to emphasize that not all proteins that bind to
antibodies qualify as antigens. For instance, immunoglobulin-binding proteins like Protein A and

Protein G, expressed by Staphylococcus aureus and Streptococcal species, are commonly
employed for antibody purification procedures [19]. Although protein A may bind to the Fab
region of antibodies, the interacting amino acids might be located in the framework region (FR)
rather than the complementarity-determining regions (CDRs) [20]. Such proteins are often
incorrectly identified as antigen. Consequently, structures exhibiting such characteristics require
meticulous verification and should be excluded from our dataset.

2.2 PDB splitting

we directly incorporated the PDB entry of antigen-antibody complex with only one antigen
and one antibody into the AACDB database. However, quite a few antigen-antibody complexes
contain several antigens and antibodies. In such cases, before splitting the structures, we need to
determine the correct pairing of light and heavy chains by examining the information on
equivalent chain interactions in the PDBsum database [21]. Next, we assign the correct antigen
chains to the antibodies by identifying whether the remaining chains in the structure interact with
the CDR regions of determined antibodies. For example, in the case of 1AHW, which contains
two copies of antigen-antibody complexes, chains BA and DE are identified as two antibody pairs,
where chains C and F bind to BA and DE, respectively. Consequently, 1AHW is split into two
files, BAC and DEF (Figure 2A). In 6OGE, Pertuzumab (chains C and B) and Trastuzumab
(chains E and D) bind to different epitopes of the receptor tyrosine-protein kinase erbB-2 (chain
A). This would generate two records in AACDB (CBA and EDA) (Figure 2B). Specifically, it is
necessary to determine if anti-idiotype antibody binds to both the heavy and light chains of the
antibody on the other side. For instance, in 3BQU, the anti-idiotype 3H6 Fab (chains DC) only
interact with the heavy chain (chain B) of 2F5 Fab. In AACDB, 2F5 Fab will be split and the light
chain (chain A) will be discarded(Figure 2C).

For each PDB entry, we utilized the corresponding split chains information to divide the
downloaded .pdb or .cif files based on the ATOM records. Since the Naccess software does not
support .cif files as input, we converted the .cif files into PDB format files while performing the
splitting process (**see Interacting residues definition**). Furthermore, we adjusted the annotation
of the split .fasta files to ensure coherence with AACDB records.

2.3 Metadata

AACDB provides detailed metadata for each entry, including chain IDs, antibody name,
antigen name, method, resolution, organism, and more. To ensure data accuracy, we have
conducted comprehensive verification by consulting original literature sources. AACDB has
addressed many annotation errors identified within the corresponding PDB entries. These errors
include but not limited to: (1) mislabeling of species (e.g. the entry 7WRL where the organism of
BD55-1239H antibody was erroneously labeled as "SARS coronavirus B012"); (2) resolution
annotation errors (e.g. 1NSN, the resolution should be 2.9 , but it was incorrectly labeled as 2.8);
(3) mislabeling of antibody chains as other proteins (e.g. in 3KS0, the light chain of B2B4
antibody was misnamed as heme domain of flavocytochrome b2); (4) misidentification of heavy
chains as light chains (e.g. both two chains of antibody were labeled as light chain in 5EBW); (5)
mutation status annotation errors. We have identified cases in which PDB entries indicate "NO"
for mutations, while in reality, mutations exist (e.g. bevacizumab (Avastin) in 6BFT was labeled
as none mutation. When aligned with the bevacizumab sequence, however, mutation T8D/T30D in
heavy chain and S52D/S53D in light chain were observed.); and (6) incomplete annotations.
Certain entries only provide the name of the mutant without specifying the precise mutation site

(e.g. in 7SU1, antibody was described as   Ipilimumab variant Ipi.106. but PDB database do not
provide any mutation amino acid or position, which can be found in the reference literature). We
carefully checked each entry manually to find out all possible annotation problems. In the
AACDB database, we have supplemented the available information or correct the annotations
through comprehensive literature reviews and sequence alignment with wild-type proteins.

The antibody nomenclature follows the title of the corresponding search entry in the RCSB
PDB database, with verification done through the original literature. In cases where the name in
the original literature differs from that in the RCSB PDB, we used the name in the published
literature as the standard. Furthermore, for antibody fragments lacking names in both the RCSB
PDB database and original literature, we adopt a naming convention of "PDBID" + "antibody
fragment" (e.g., 4WEB Fab).

The biological and physicochemical properties play a crucial role in the new drug discovery
pipelines for therapeutic antibodies. These properties include solubility, immunogenicity,
aggregation tendencies, expression level, stability, and hydrophobicity. We provide the
International Non-proprietary Name (INN) and the clinical trial for each therapeutic antibody
entry, linking them to the DOTAD database [22]. Numerous antigens have been successfully
identified as targets for antibodies or small molecule drugs. We conducted a comparison between
antigen sequences and the drug targets listed in the DrugBank database [23]. A threshold of
percent identity > 90% was applied to determine the corresponding drug targets.

2.4 Interacting residues definition

We labeled the interacting residues based on SASA and atom distance. Naccess V2.1.1 and
Bio.PDB module were employed to calculate SASA value for each residue in antibody and

antigen, respectively. The residues with a SASA loss (ΔSASA) in binding of more than 1Å^2 were

classified as interacting residues. In addition, we also defined another set of interacting
paratope-epitope residues by a distance cutoff of 5Å . Two amino acids are considered as
interacting residues if they have at least one atom within a distance of 5 Å from any atom.

2.5 Data integration and website implementation

The main data processing algorithm is implemented in Python. The front-end web interface
of AACDB was constructed by HTML and enhanced with JavaScript, CSS and Bootstrap
technologies. We developed a dynamic 3D structure visualization window based on PV, a
WebGL-based protein viewer, inspiring by Dunbar *et al.* [15]. All the data were managed within
the MySQL database system. For the back-end functionality, PHP is utilized to enable data
browsing, searching, and downloading features.

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3.   Results

3.1 Statistics

Out of more than 32000 experimental structures in the PDB database, 7498 antigen−antibody
entries were manually curated in current version of AACDB, referring to 16 antibody fragment
types across 14 species(Figure 3). It is obvious that fab fragments and human antibodies
accounted for the largest proportion of the data, accounting for 71.98% and 60.95%, respectively.
Our statistical analysis reveals a significant increase in the number of antigen-antibody complex
entries within the PDB during the period from 2021 to 2023. These entries accounted for
approximately 45% of the total antigen-antibody complex entries. The data in Figure 2B, released

in 2024, reflects a change due to the removal of the 7SIX entry (initially released on November 16,
2022) from the distribution of released PDB entries. This removal occurred on January 17, 2024
and it has been replaced by the entry 8TM1. Furthermore, the developability properties of
antibodies in 325 entries can be queried in the DOTAD database, at the meanwhile, 3,733 antigen
records have been identified as drug targets in DrugBank (data not shown).

3.2 Database browse and search

All the data can be browsed directly by clicking the "Datasets" item on the top menu (Figure
4). The summary table includes nine columns as following:

(1) AACDB_ID: The unique id in the AACDB database, linking to the "Detail" page;

(2) PDBID: The identifier of the RCSB PDB database (https://www.rcsb.org/), linking to PDB;

(3) Chains: The chain ids containing in this entry. Antibody and antigen chains were separated by
"_". The heavy chain id precedes the light chain id if antibody is complete;

(4) Antibody: This column is represented by "antibody name + fragment";

(5) Protein: The corresponding antigen;

(6) Organism: The source organism of antibody and antigen;

(7) Method: The experiment method used to solve the structure;

(8) Resolution: The indicator that measures the resolution of protein structures in experiments is
expressed in units of angstroms (Å).

(9) Reference: The DOI linker of original literature that produce this structure.

The table in AACDB can be easily searched through the search panel at the top section of the
"Datasets" page. Users can perform a quick search by specifying one or more fields and entering
relevant keywords. The search results can be downloaded as files in either txt or csv format.
Figure 4 shows the search results using the keyword "Lysozyme" in "Protein" column, returning
117 hits.

An individual structure can be accessed using its AACDB_ID accession code. When click the
AACDB_ID hyperlink, the user will be brought to its details page as shown in Figure 5. There, the
complex structure can be visualized with different colors and styles (Figure 5A). Besides the
visualization window, we provide more details of this entry, including the mutation, INN and
clinical trial of antibody, the ID in DrugBank of antigen (if exist) (Figure 5B). Under the structure
information tab, further details about each chain can be found. These include: sequence, mutate
amino acid type and position, interacting residues in each chain based on the ΔSASA method,
interacting plot of paratope-epitope residues by a distance cutoff of < 5Å (Figure 5D).

3.3 Data download

We provide two ways for downloading the data:

(1) Download data of the single entry.

When access the detailed information about an entry using the corresponding AACDB_ID,
user can click the hyperlink at the "Download" section of the bottom of the page to download the
data for a single certain entry (Figure 5E).

(2) Download all the data of AACDB.

AACDB provide the download page for users. All the sequence and structure files and the
interacting data based on different methods were packaged in different .zip file that can be
downloaded.

Moreover, the website provides a user-friendly 'Help' page that presents a step-by-step
tutorial to assist users in manipulating, querying, browsing and downloading the AACDB

database.

4. Discussion and conclusion

Research on antigen-antibody interactions contributes to the advancement of the antibody-related industry. Databases such as PDB and SabDab provide the foundational data for this purpose. However, there are still many unresolved issues. In this work, we have developed the AACDB database with the aim of providing a clean and reliable dataset of antigen-antibody interactions. During the process of data collection and organization, we identified numerous annotation errors in the PDB database. Some of these errors had been directly introduced into SabDab. For example, the species of the antibody in 7WRL was incorrectly labeled as "SARS coronavirus B012" in both PDB and SabDab. We also invested significant effort and time to manually cross-reference with original literature in order to rectify these errors and exclude antibody binding proteins that were erroneously annotated as antigens by SabDab. Apart from the curation and reannotation of structural data, AACDB offers features not provided by other antigen-antibody complex databases: 1) AACDB's data processing pipeline supports mmCIF files, and 2) we provide amino acids in the interaction interface through two methods, enabling the definition of unified standards for epitopes and paratopes. This provides a more accurate and comprehensive benchmark dataset for developed interaction interface prediction tools, enhancing the comparability of various tools.

However, AACDB still has some limitations. Despite our best efforts, the limitations of our team's resources and knowledge mean that our database may not capture all antigen-antibody complex structures. Since the data is manually curated, eliminating errors completely during the information processing is a challenge. While we strive to fill in any gaps, we also hope that experts and users within the community can provide timely feedback to help us improve these issues. Additionally, currently AACDB only includes antigen proteins with a length greater than 50 amino acids. In future work, we will expand the database to include complex structures of different antigen types, including peptides, nucleic acids, and haptens.

In summary, AACDB is a novel database of antigen-antibody complexes that provides information on antibody-developability, antigen-drug target relationships, and detailed antigen-antibody interaction interfaces. It is fully accessible at http://i.uestc.edu.cn/AACDB. We are committed to a regular data update, ensuring that researchers in immunoinformatics have access to timely and valuable resources.

5. Acknowledgment(s)

This work was supported by the National Natural Science Foundation of China [grant numbers 62071099, 62371112] and Sichuan Province Science and Technology Support Program [2024NSFSC0636].

6. CRediT author statement

Yuwei Zhou: Writing - Original Draft, Software, Conceptualization **Wenwen Liu:** Validation, Visualization **Ziru Huang:** Validation **Yushu Gou:** Data Curation **Siqi Liu:** Data Curation **Lixu Jiang:** Data Curation **Yue Yang:** Data Curation **Jian Huang:** Supervision, Funding acquisition

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Figure 1. Data processing algorithm pipeline of AACDB.

Figure 2. Examples of PDB file splitting under different situations. (A) 1AHW contains two copies of the same antigen and the same antibody. (B) In 6OGE, two different antibodies bind to distinct epitopes of the same antigen. (C) In 3BQU, an anti-idiotypic antibody binds exclusively to a single chain of the antibody.

Figure 3. AACDB statistics. (A) Antibody fragment distribution in database. (B) The number of antibody-antigen complexes released in different year (unique PDBID). (C) Organismal distribution of antibody entries.

Figure 4. The AACDB browse and search page. This example demonstrates the use of the keyword 'Lysozyme' in the 'Protein' column, showcasing the search functionality of the AACDB database. The search results display antigen related to Lysozyme.

Figure 5. The details page of entries, taking '1BJ1' as an example. (A) Structure visualization window. (B) Entry meta information. (C) sequence and mutation information. (D) Interacting residues details based on SASA and atom distance methods. (E) The download hyperlinks of a single entry.

A

1AHW → **1AHW_BA_C** **1AHW_ED_F**

5G9 Heavy chain   5G9 Light chain   Tissue factor

B

6OGE → **6OGE_CB_A** **6OGE _ED_A**

Pertuzumab Light chain   Trastuzumab Light chain   Receptor tyrosine - protein kinase erbB-2
Pertuzumab Heavy chain   Trastuzumab Heavy chain

C

3BQU → **3BQU_DC_B**

anti-idiotype 3H6 Light chain   2F5 Fab Light chain
anti-idiotype 3H6 Heavy chain   2F5 Fab Heavy chain

Datasets

Search

PDBID: [   ]   Antibody: [   ]   Protein: [Lysozyme]

Organism: [   ]   Resolution(<): [   ]   All fields: [   ]

[Search]

Submit search →

Click to download the search results ↘

Results list: **117** hits (Download the results:

 or 📄)

Click for the details

Click to get original literature ↙

AACDB_ID	PDBID	Chains	Antibody	Protein	Organism	Method	Resolution	Reference
1	1A2Y	BA_C	D1.3 Fv	Lysozyme C	Mus musculus, Gallus gallus	XRD	1.50	10.1021/bi980148j
2	1BQL	HL_Y	HYHEL-5 Fab	BOBWHITE QUAIL Lysozyme	Mus musculus, Colinus virginianus	XRD	2.60	10.1002/(SICI)1097-0134(199609)26:1<55::AID-PROT5>3.0.CO;2-F
3	1BVK	BA_C	HULYS11 Fv	Lysozyme	Homo sapiens, Gallus gallus	XRD	2.70	10.1084/jem.187.4.479
4	1BVK	ED_F	HULYS11 Fv	Lysozyme	Homo sapiens, Gallus gallus	XRD	2.70	10.1084/jem.187.4.479
5	1C08	BA_C	HYHEL-10 Fv	Hen Egg White Lysozyme	Mus musculus, Gallus gallus	XRD	2.30	10.1074/jbc.274.39.27623
6	1DQJ	BA_C	HyHEL-63 Fab	Lysozyme C	Mus musculus, Gallus gallus	XRD	2.00	10.1021/bi000054l
7	1DZB	A_X	1F9 scFv	Turkey Egg White Lysozyme	Mus musculus, Meleagris gallopavo	XRD	2.00	10.1006/jmbi.2000.3971
8	1DZB	B_Y	1F9 scFv	Turkey Egg White Lysozyme	Mus musculus, Meleagris gallopavo	XRD	2.00	10.1006/jmbi.2000.3971
9	1FBI	HL_X	F9.13.7 Fab	GUINEA FOWL Lysozyme	Mus musculus, Numida meleagris	XRD	3.00	10.1074/jbc.270.30.18067
10	1FBI	QP_Y	F9.13.7 Fab	GUINEA FOWL	Mus musculus, Numida	XRD	3.00	10.1074/jbc.270.30.18067

Click to view the entry in PDB database

A

Structure visualisation

H.TYR102 [Heavy Chain]

Heavy Chain
Light Chain
Antigen Chain

Display styles :
○ Spacefill
○ Wire
○ Ball&stick
● Cartoon

Display color :
● Default
○ Color by B-factor
○ Color by sec.structure
○ Color by element

□ Spin on/off

B

Entry information

Complex	
PDBID:	1BJ1
Chains:	HL_W
Organism:	Mus musculus, Homo sapiens
Method:	XRD
Resolution (Å):	2.40
Reference:	10.1016/s0969-2126(98)00116-6
Antibody	
Antibody:	12 Fab
Antibody mutation:	No
INN (Clinical Trial):	Bevacizumab(Approved); Dilpacimab(Phase-II); Ivonescimab(Preregistration);
Antigen	
Antigen:	Vascular endothelial growth factor A
Antigen mutation:	No
Durg Target:	P15692

Click to access the developability information in DOTAD database

Click to access the target information in DurgBnak database

C

Sequence information

Antibody

Heavy Chain: H
Mutation: NULL

>1BJ1_H|Chain B[auth H], F[auth K]|Fab fragment, heavy chain|Mus musculus (10090)
EVQLVESGGGLVQPGGSLRLSCAASGYTFTNYGMNWVRQAPGKGLEWVGWINTYTGEPTYAADFKRRFTFSLDTSKST
AYLQMNSLRAEDTAVYYCAKYPHYYGSSHWYFDVWGQGTLVTVSSASTKGPSVFPLAPSSKSTSGGTAALGCLVKDYFP
EPVTVSWNSGALTSGVHTFPAVLQSSGLYSLSSVVTVPSSSLGTQTYICNVNHKPSNTKVDKKVEPKSCDKTHT

Light Chain: L
Mutation: NULL

>1BJ1_L|Chain A[auth JJ], E[auth L]|Fab fragment, light chain|Mus musculus (10090)
DIQMTQSPSSLSASVGDRVTITCSASQDISNYLNWYQQKPGKAPKVLIYFTSSLHSGVPSRFSGSGSGTDFTLTISSLQPE
DFATYYCQQYSTVPWTFGQGTKVEIKRTVAAPSVFIFPPSDEQLKSGTASVVCLLNNFYPREAKVQWKVDNALQSGNSQ
ESVTEQDSKDSTYSLSSTLTLSKADYEKHKVYACEVTHQGLSSPVTKSFNRGEC

Antigen

Chain: W
Mutation: NULL

>1BJ1_W|Chain C[auth V], D[auth W]|Vascular endothelial growth factor A|Homo sapiens (9606)
GQNHHEVVKFMDVYQRSYCHPIETLVDIFQEYPDEIEYIFKPSCVPLMRCGGCCNDEGLECVPTEESNITMQIMRIKPHQ
GQHIGEMSFLQHNKCECRPKKD

E

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Download sequences
Download structure
Download interacting residues (ΔSASA based)
Download interacting residues (Distance based)

D

Interaction

1、**Solvent accessible surface areas (SASA)** were calculated (Naccess V2.1.1) for each residue in antibody and antigen, respectively. The residues with SASA loss in binding of more than $1Å^2$ were classified as interacting residues.

Interacting residues (ΔSASA based)

L: TYR91 SER92 THR93 VAL94 TRP96

H: THR30 ASN31 TYR32 GLY33 TRP50 ASN52 THR53 TYR54 THR55 GLU57 THR59 TYR99 PRO100 HIS101 TYR102 TYR103 GLY104 SER105 SER106 HIS107 TRP108

W: TYR45 LYS48 GLN79 ILE80 MET81 ARG82 ILE83 LYS84 HIS86 GLN87 GLY88 GLN89 HIS90 ILE91 GLY92 GLU93 MET94

2、We defined interacting paratope-epitope residues by a **distance cutoff of < 5Å**. Two amino acids are considered as interacting residues if they have at least one atom within a distance of 5 Å from any atom.

Interacting residues (Atom distance based)