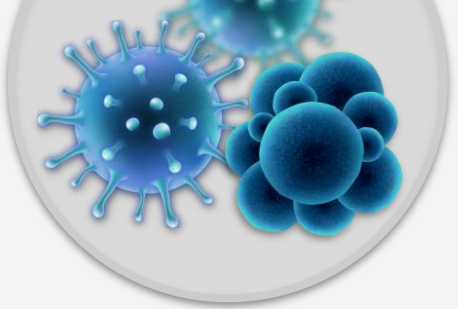
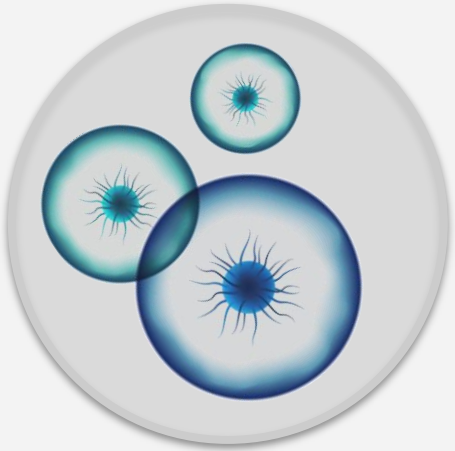
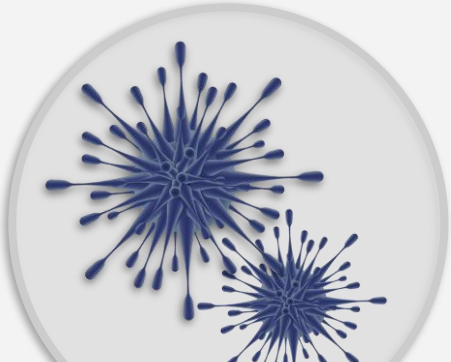
RBP-ID

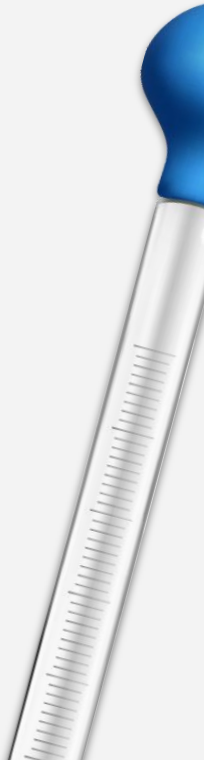
Development of a Web-Based Bioinformatics Tool for the Identification of Phage Receptor-Binding Proteins

Presented by Cátia Rosário
Oriented by Silvio Santos and Óscar Dias









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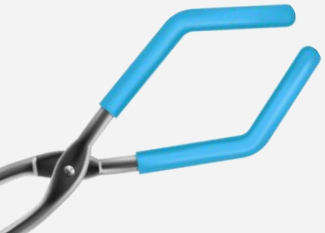
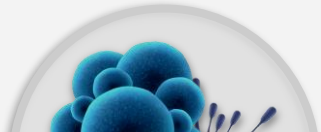
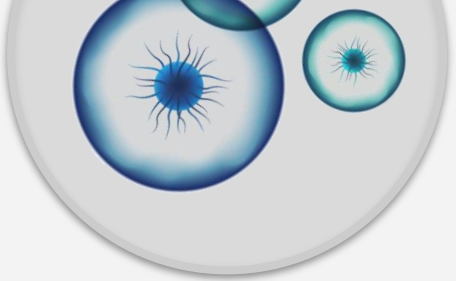
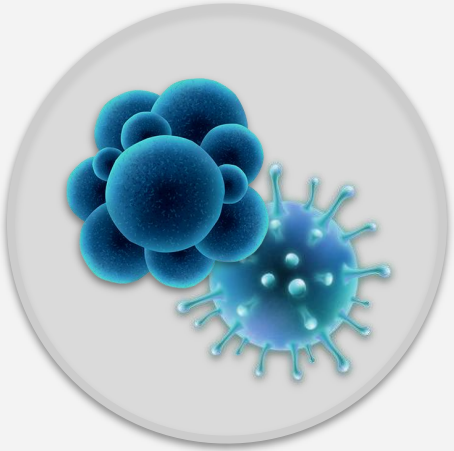
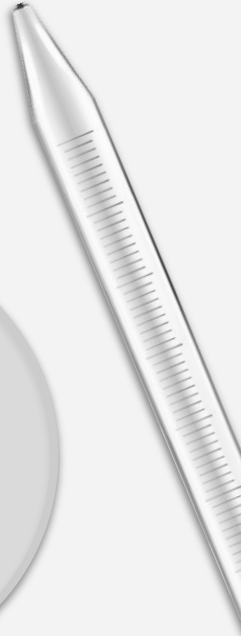








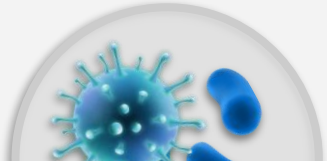


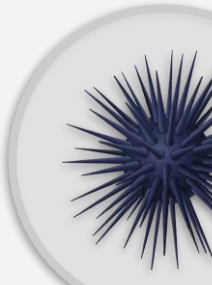
01

Milestones

Milestones reached

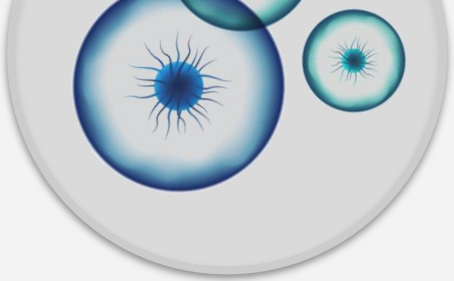
- Initial Research about the theme
- Literature Compilation about Viruses and RBPs
- 1st Dataset creation with manually compiled RBPs candidates
- RBPs Validation through published experimental data
- 2nd Dataset creation with computationally predicted RBPs from a compilation of Datasets
- 3nd Dataset creation with computationally predicted and validated RBPs
- Initial Bioinformatic Analysis – Clustering with cd-hit and K-means
- Initial Bioinformatic Analysis – PCA and Dendrograms

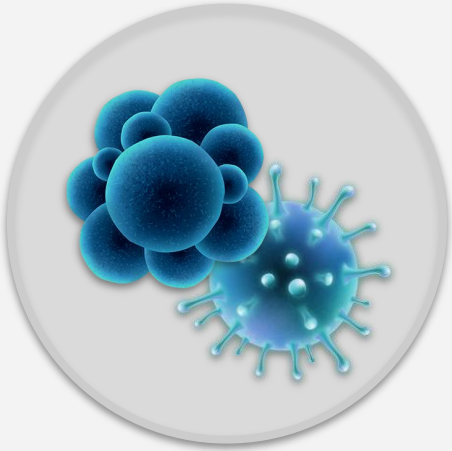


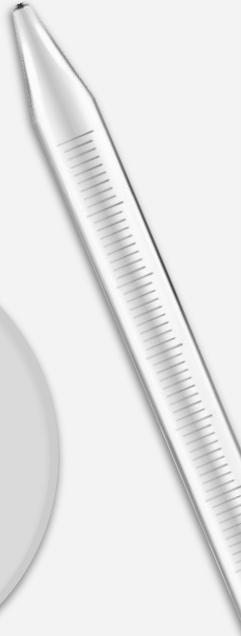


02

Abstract

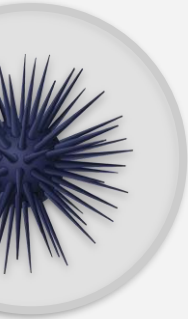


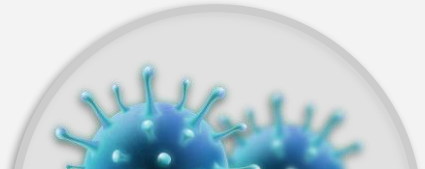


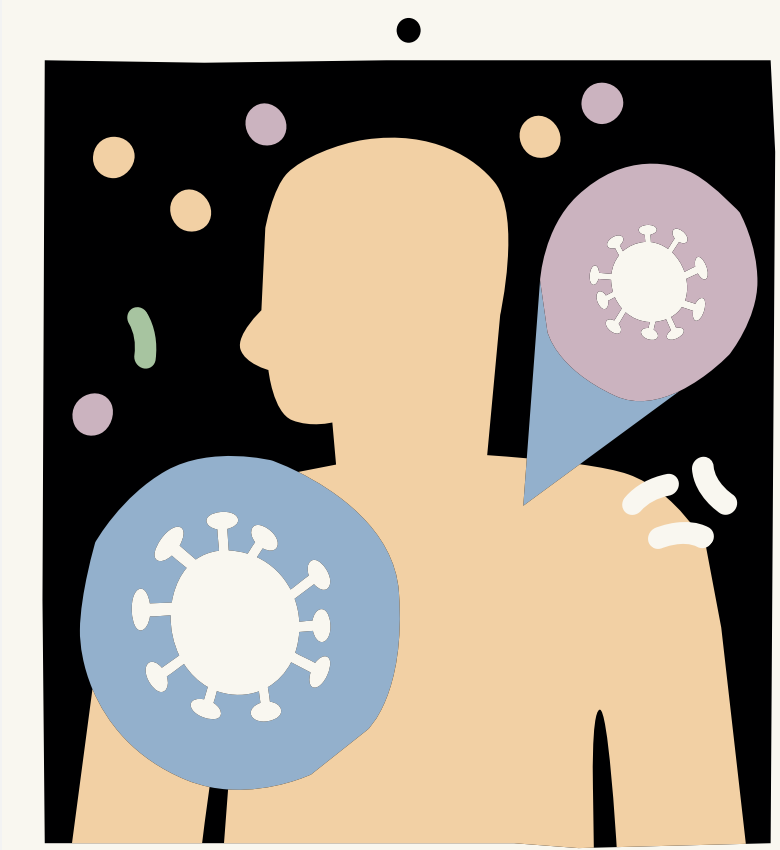


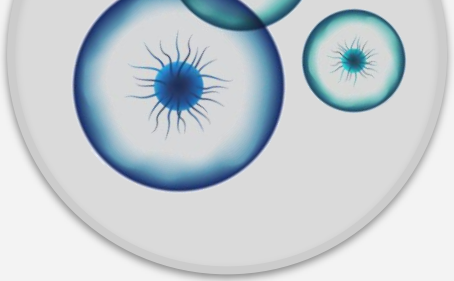
Abstract

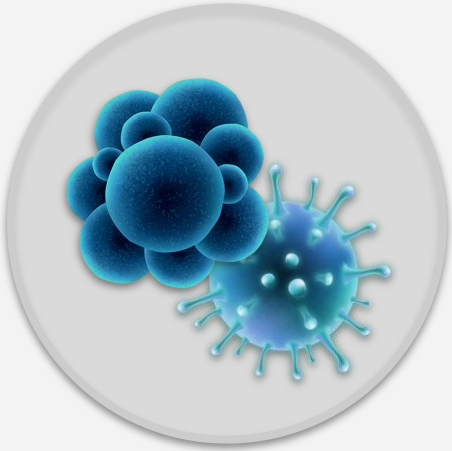
- **Bacteriophages:** Viruses that infect bacteria with key roles in biotechnology and medicine.
- **Receptor-Binding Proteins (RBPs):** Determine host specificity by mediating bacterial recognition and attachment.
- **Challenge:** RBPs exhibit high sequence diversity and low conservation, making identification difficult.
- **Solution – RBP-ID:** A web-based bioinformatics tool for systematic RBP prediction in phage genomes.

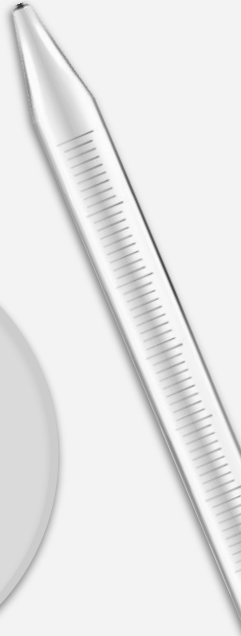












03

Introduction

Introduction

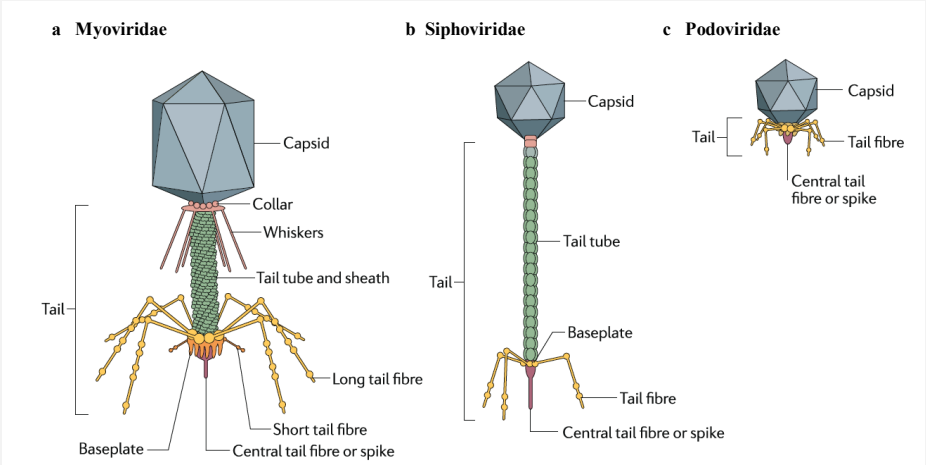
Background

Phage display & delivery systems: Technologies for gene editing and therapeutic biomolecule delivery.
Adaptation mechanisms: Co-evolution with bacteria through receptor mutations and CRISPR-Cas defence systems.
Computational tools for phage genome annotation:

- Hybrid approaches: HMM searches and ML classifiers
- Host prediction via RBP analysis: Random Forest and SVMs, XGBoost
- PHYPred: Functional annotation tools for phage-encoded enzymes
- PhANNs

Importance

Biotechnological & therapeutic: Antibiotic resistance solutions, microbiome engineering, biosensing and precision medicine
Phages vs. antibiotics: Target specific pathogens without harming beneficial bacteria, a key advantage in antimicrobial resistance (AMR)
Synthetic biology advances: Engineered phages and phage-derived enzymes broaden applications in medicine, agriculture, and industry
Need for advanced bioinformatics tools



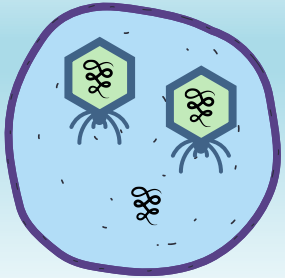
Morphotypes

Tail Fibre World

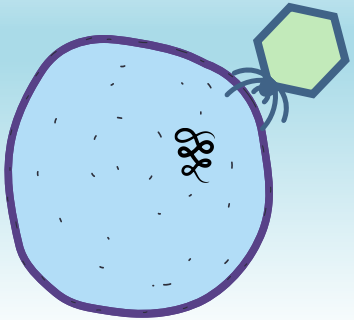
Short, Long, Spikey, Contractile, Non-Contractile

HOW DO VIRUSES REPLICATE

The virus invades the host cell

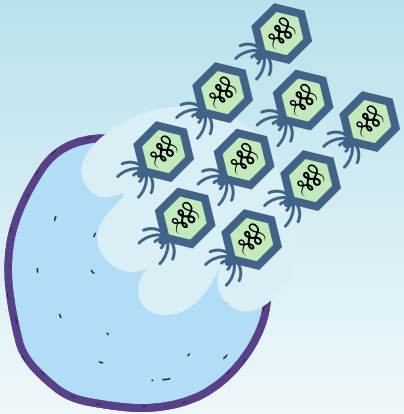


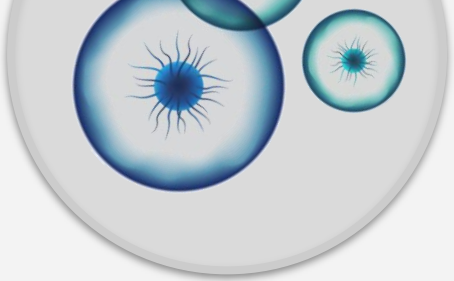
The virus's genetic material reprograms the host's DNA and starts to replicate itself

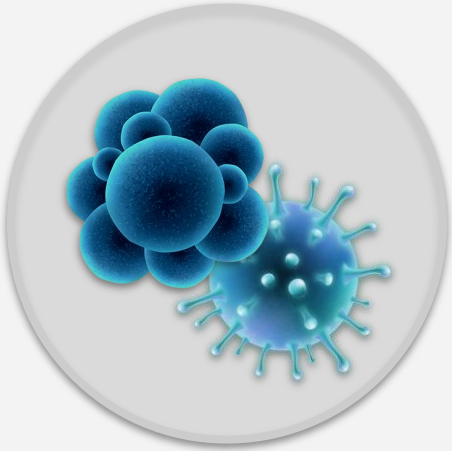


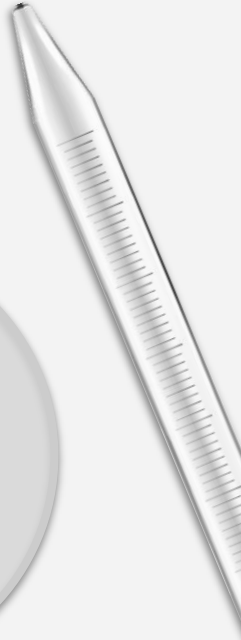
The virus injects its genetic material into the host cell

When the host cell is filled with virus copies, it bursts, releasing them into the bloodstream to repeat the process



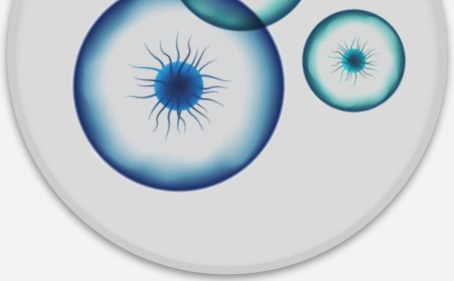


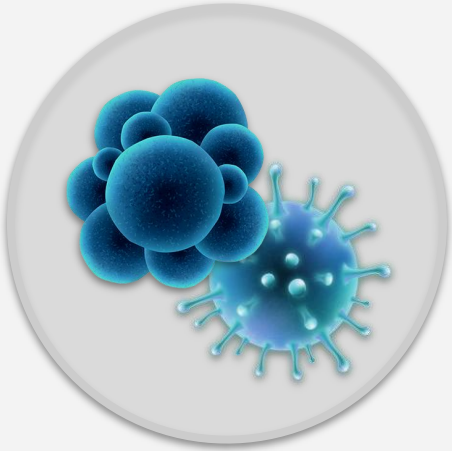


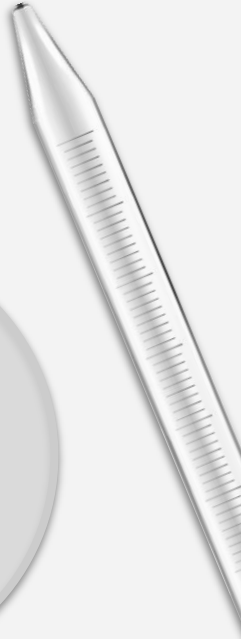


04

Results

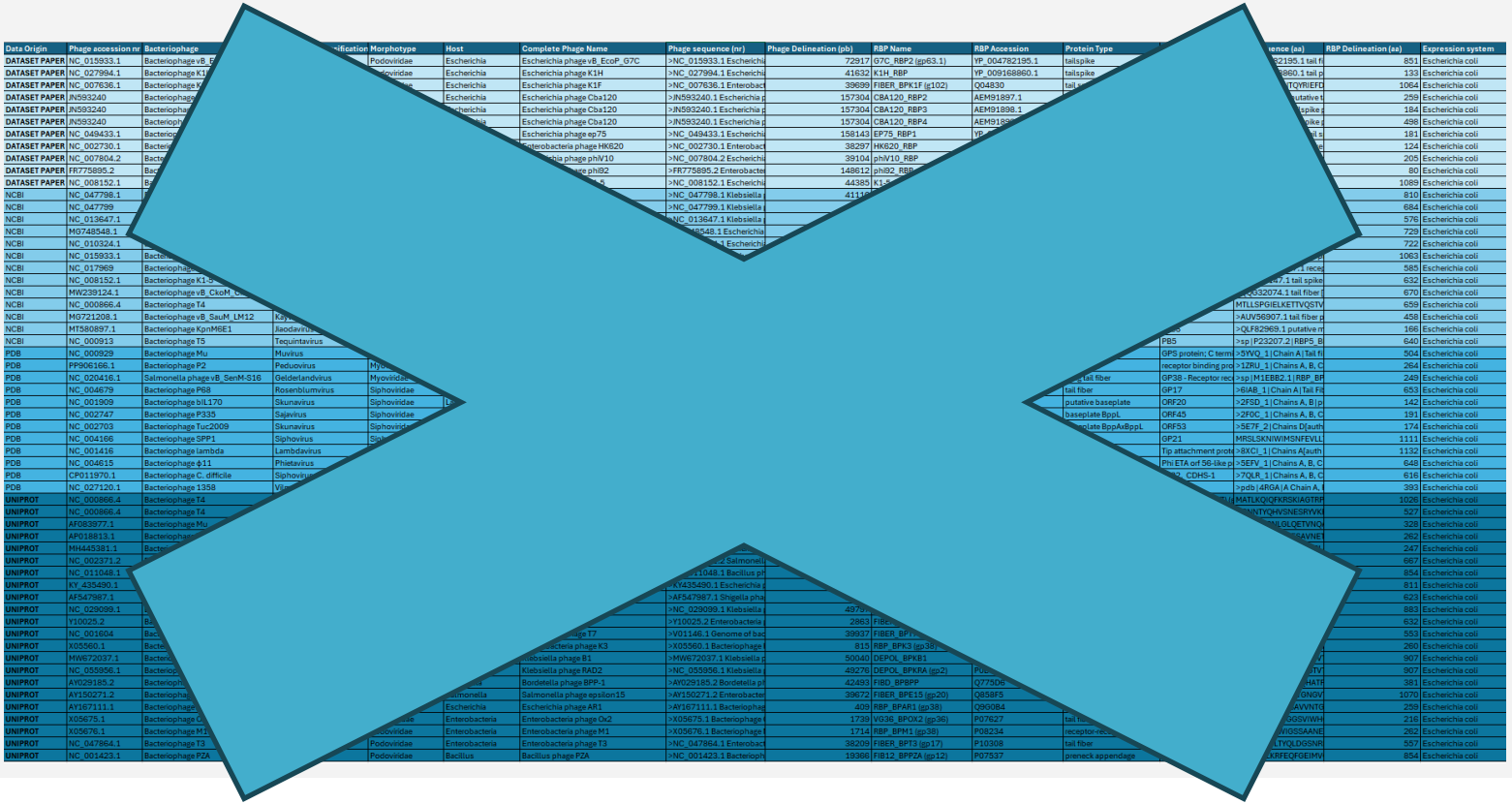






4.1

1st Dataset



Data Origin	Phage accession nr	Bacteriophage		Morphotype	Host	Complete Phage Name	Phage sequence (nr)	Phage Delineation (pb)	RBP Name	RBP Accession	Protein Type		Delineation (aa)	Expression system
DATASET PAPER	NC_015933.1	Bacteriophage vB_E		Podoviridae	Escherichia	Escherichia phage vB_EcoP_G7C	>NC_015933.1 Escherichi	72917	G7C_RBP2 (gp63.1)	YP_004782195.1	tailspike		851	Escherichia coli
DATASET PAPER	NC_027994.1	Bacteriophage K1		oviridae	Escherichia	Escherichia phage K1H	>NC_027994.1 Escherichi	41632	K1H_RBP	YP_009168860.1	tailspike		133	Escherichia coli
DATASET PAPER	NC_007636.1	Bacteriophage K		us	Escherichia	Escherichia phage K1F	>NC_007636.1 Enterobact	39699	FIBER_BPK1F (gp102)	Q04830	tail K		1064	Escherichia coli
DATASET PAPER	JN593240	Bacteriophage			Escherichia	Escherichia phage Cba120	>JN593240.1 Escherichia p	157304	CBA120_RBP2	AEM91897.1			259	Escherichia coli
DATASET PAPER	JN593240	Bacteriophage			herichia	Escherichia phage Cba120	>JN593240.1 Escherichia p	157304	CBA120_RBP3	AEM91898.1			184	Escherichia coli
DATASET PAPER	JN593240	Bacteriophage				Escherichia phage Cba120	>JN593240.1 Escherichia p	157304	CBA120_RBP4	AEM91899			498	Escherichia coli
DATASET PAPER	NC_049433.1	Bacteriop				Escherichia phage ep75	>NC_049433.1 Escherichi	158143	EP75_RBP1	YP_			181	Escherichia coli
DATASET PAPER	NC_002730.1	Bacteri				Enterobacteria phage HK620	>NC_002730.1 Enterobact	38297	HK620_RBP				124	Escherichia coli
DATASET PAPER	NC_007804.2	Bacte				phage phiV10	>NC_007804.2 Escherichi	39104	phiV10_RBP				205	Escherichia coli
DATASET PAPER	FR775895.2	Bac				phi82	>FR775895.2 Enterobacter	148612	phi82_RBP				80	Escherichia coli
DATASET PAPER	NC_008152.1	Ba				5	>NC_008152.1 Escherichi	44385	K1-5				1089	Escherichia coli
NCBI	NC_047798.1						>NC_047798.1 Klebsiella	41119					810	Escherichia coli
NCBI	NC_047799						>NC_047799.1 Klebsiella						684	Escherichia coli
NCBI	NC_013647.1						>NC_013647.1 Klebsiella						576	Escherichia coli
NCBI	MG748548.1						8548.1 Escherichia						729	Escherichia coli
NCBI	NC_010324.1						1 Escherichi						722	Escherichia coli
NCBI	NC_015933.1	Bacte											1063	Escherichia coli
NCBI	NC_017969	Bacteriophage											585	Escherichia coli
NCBI	NC_008152.1	Bacteriophage K1-5											632	Escherichia coli
NCBI	MW239124.1	Bacteriophage vB_CkoM_C											670	Escherichia coli
NCBI	NC_000866.4	Bacteriophage T4											659	Escherichia coli
NCBI	MG721208.1	Bacteriophage vB_SauM_LM12	Kay										458	Escherichia coli
NCBI	MT580897.1	Bacteriophage KpnM6E1	Jiaodavirus										166	Escherichia coli
NCBI	NC_000913	Bacteriophage T5	Tequintavirus								PB5	>sp |P23207.2| RBP5_B	640	Escherichia coli
PDB	NC_000929	Bacteriophage Mu	Muvirus								GPS protein; C term	>5YVQ_1 | Chain A | Tail fi	504	Escherichia coli
PDB	PP906166.1	Bacteriophage P2	Peduovirus	Myo							receptor binding pro	>1ZRU_1 | Chains A, B, C	264	Escherichia coli
PDB	NC_020416.1	Salmonella phage vB_SenM-S16	Geldertandvirus	Myoviridae							GP38 - Receptor rec	>sp |M1E8B2.1 | RBP_BF	249	Escherichia coli
PDB	NC_004679	Bacteriophage P68	Rosenblumvirus	Siphoviridae							GP17	>6IAB_1 | Chain A | Tail Fi	653	Escherichia coli
PDB	NC_001909	Bacteriophage bIL170	Skunavirus	Siphoviridae							ORF20	>2F5D_1 | Chains A, B | p	142	Escherichia coli
PDB	NC_002747	Bacteriophage P335	Sajavirus	Siphoviridae							ORF45	>2F0C_1 | Chains A, B, C	191	Escherichia coli
PDB	NC_002703	Bacteriophage Tuc2009	Skunavirus	Siphoviridae							ORF53	>5E7F_2 | Chains D[auth	174	Escherichia coli
PDB	NC_004166	Bacteriophage SPP1	Siphovirus	Sip							GP21	MRSLSKNIWIMSNFEVLL	1111	Escherichia coli
PDB	NC_001416	Bacteriophage lambda	Lambdavirus								Tip attachment prote	>8XCI_1 | Chains A[auth	1132	Escherichia coli
PDB	NC_004615	Bacteriophage ϕ11	Phietavirus								Phi ETA orf 56-like p	>5EFV_1 | Chains A, B, C	648	Escherichia coli
PDB	CP011970.1	Bacteriophage C. difficile	Siphovirus								2_CDHS-1	>7QLR_1 | Chains A, B, C	616	Escherichia coli
PDB	NC_027120.1	Bacteriophage 1358	Vi									>pdb |4RGA| A Chain A,	393	Escherichia coli
UNIPROT	NC_000866.4	Bacteriophage T4										MATLKQIQFKFSKIAGTRF	1026	Escherichia coli
UNIPROT	NC_000866.4	Bacteriophage T4										NTYQHVSNSESRYVK	527	Escherichia coli
UNIPROT	AF083977.1	Bacteriophage Mu										MLQLQETVNQ	328	Escherichia coli
UNIPROT	AP018813.1	Bacteriophage										SSAVNET	262	Escherichia coli
UNIPROT	MH445381.1	Bacte											247	Escherichia coli
UNIPROT	NC_002371.2					2 Salmonell							667	Escherichia coli
UNIPROT	NC_011048.1					11048.1 Bacillus ph							854	Escherichia coli
UNIPROT	KY_435490.1					KY435490.1 Escherichia							811	Escherichia coli
UNIPROT	AF547987.1					>AF547987.1 Shigella pha							623	Escherichia coli
UNIPROT	NC_029099.1	Ba				>NC_029099.1 Klebsiella		4979					883	Escherichia coli
UNIPROT	Y10025.2	Ba				>Y10025.2 Enterobacteria		2863	FIBer				632	Escherichia coli
UNIPROT	NC_001604	Bac				age T7	>V01146.1 Genome of bac	39937	FIBER_BP1				553	Escherichia coli
UNIPROT	X05560.1	Bacte				acteria phage K3	>X05560.1 Bacteriophage	815	RBP_BPK3 (gp38)				260	Escherichia coli
UNIPROT	MW672037.1	Bacte				bsiella phage B1	>MW672037.1 Klebsiella p	50040	DEPOL_BPKB1				907	Escherichia coli
UNIPROT	NC_055956.1	Bacterio				Klebsiella phage RAD2	>NC_055956.1 Klebsiella	49276	DEPOL_BPKRA (gp2)	P0			907	Escherichia coli
UNIPROT	AY029185.2	Bacterioph				Bordetella phage BPP-1	>AY029185.2 Bordetella ph	42493	FIBD_BPBPP	Q775D6			381	Escherichia coli
UNIPROT	AY150271.2	Bacteriopha			almonella	Salmonella phage epsilon15	>AY150271.2 Enterobacter	39672	FIBER_BPE15 (gp20)	Q858F5			1070	Escherichia coli
UNIPROT	AY167111.1	Bacteriophage			Escherichia	Escherichia phage AR1	>AY167111.1 Bacteriophag	409	RBP_BPAR1 (gp38)	Q90084			259	Escherichia coli
UNIPROT	X05675.1	Bacteriophage O			Enterobacteria	Enterobacteria phage Ox2	>X05675.1 Bacteriophage	1739	VG36_BPOX2 (gp36)	P07627	tail fi		216	Escherichia coli
UNIPROT	X05676.1	Bacteriophage M1			Enterobacteria	Enterobacteria phage M1	>X05676.1 Bacteriophage	1714	RBP_BPM1 (gp38)	P08234	receptor-recog		262	Escherichia coli
UNIPROT	NC_047864.1	Bacteriophage T3		Podoviridae	Enterobacteria	Enterobacteria phage T3	>NC_047864.1 Enterobact	38209	FIBER_BPT3 (gp17)	P10308	tail fiber		557	Escherichia coli
UNIPROT	NC_001423.1	Bacteriophage PZA		Podoviridae	Bacillus	Bacillus phage PZA	>NC_001423.1 Bacterioph	19366	FIB12_BPPZA (gp12)	P07537	preneck appendage		854	Escherichia coli

1st Dataset – Validated RBPs

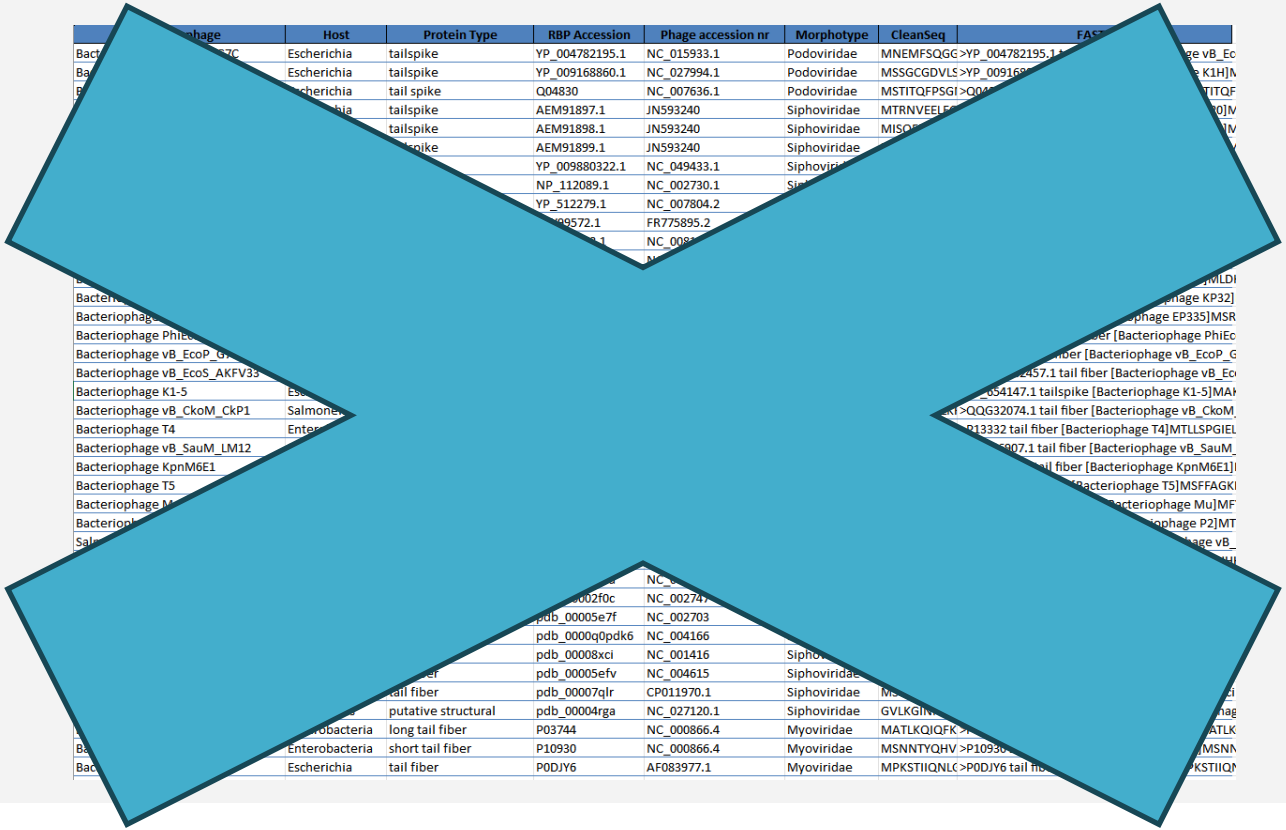
```python
import pandas as pd
from Bio import Entrez, SeqIO
import time

Entrez.email = "catiarosario10@gmail.com"
excel_path = "RBP_validated_final.xlsx"
sheet_name = "Sheet1"
phage_id_col = "Bacteriophage"
protein_id_col = "RBP Accession"
desc_col = "Protein Type"
source_col = "Bacteriophage"
seq_col = "RBP sequence (aa)"
output_fasta_file = "validated_RBP.fasta"
output_excel_file = "validated_processed.xlsx"

def clean_sequence(seq):
    if not isinstance(seq, str):
        return ""
    lines = seq.strip().splitlines()
    lines = [line for line in lines if not line.startswith(">")]
    return "".join(lines).replace(" ", "").upper()

def make_fasta_header(row):
    desc = row.get(desc_col) if pd.notna(row.get(desc_col)) else ""
    source = row.get(source_col) if pd.notna(row.get(source_col)) else ""
    return f">{row[protein_id_col]} {desc} [{source}]".strip()

def make_full_fasta(row):
    header = make_fasta_header(row)
    seq = row["CleanSeq"]
    return f"{header}\n{seq}"

def get_taxid_and_accession(phage_name):
    try:
        search = Entrez.esearch(db="nucleotide", term=phage_name, retmode="xml")
        record = Entrez.read(search)
        search.close()
        if not record["IdList"]:
            return None, None
        accession_id = record["IdList"][0]
        handle = Entrez.efetch(db="nucleotide", id=accession_id, rettype="gb", retmode="text")
        seq_record = SeqIO.read(handle, "genbank")
        handle.close()
        taxid = None
        for feature in seq_record.features:
            if feature.type == "source":
                for ref in feature.qualifiers.get("db_xref", []):
                    if ref.startswith("taxon:"):
                        taxid = ref.split(":")[1]
                        break
        return taxid, accession_id
    except Exception as e:
        print(f"[{phage_name}] TaxID/Accession fetch error: {e}")
        return None, None
```

```python
df = pd.read_excel(excel_path, sheet_name=sheet_name)
df = df.dropna(subset=[phage_id_col, protein_id_col, seq_col])

df["CleanSeq"] = df[seq_col].apply(clean_sequence)

df["FASTA"] = df.apply(make_full_fasta, axis=1)

seq_to_phageids = {}
for _, row in df.iterrows():
    seq = row["CleanSeq"]
    phageid = row[phage_id_col]
    seq_to_phageids.setdefault(seq, []).append(phageid)

def get_duplicate_phageids(row):
    seq = row["CleanSeq"]
    phageid = row[phage_id_col]
    duplicates = [pid for pid in seq_to_phageids.get(seq, []) if pid != phageid]
    return ", ".join(duplicates) if duplicates else ""

df["DuplicatePhageIDs"] = df.apply(get_duplicate_phageids, axis=1)

df_unique = df.drop_duplicates(subset=["CleanSeq"]).copy()

required_cols = ["Bacteriophage", "Phage accession nr", "Protein Type", "Host", "RBP Accession", "Morphotype"]
missing_cols = [col for col in required_cols if col not in df_unique.columns]
if missing_cols:
    raise ValueError(f"Missing columns: {missing_cols}")

df_unique["AccessionFetched"] = df_unique["Phage accession nr"]
df_unique["Protein Type"] = df_unique["Protein Type"]
df_unique["Bacteriophage"] = df_unique["Bacteriophage"]

with open(output_fasta_file, "w", encoding="utf-8") as fasta_out:
    for fasta_entry in df_unique["FASTA"]:
        fasta_out.write(fasta_entry + "\n")

cols_to_save = ["Bacteriophage", "Host", "Protein Type", "RBP Accession", "Phage accession nr", "Morphotype", "CleanSeq", "FASTA"]

df_unique.to_excel(output_excel_file, columns=cols_to_save, index=False)
print(f"FASTA saved to '{output_fasta_file}'")
print(f"Excel saved to '{output_excel_file}'")
```

Workflow data processing - VALIDATED:

- Add FASTA format when missing
- Create a column with only aa to remove duplicates
- Create column with FASTA format to create FASTA file
- Create column with duplicated identifiers when found
- Remove duplicate lines based on sequence
- Create FASTA file and Excel file

Validated RBPs – Processement with Python

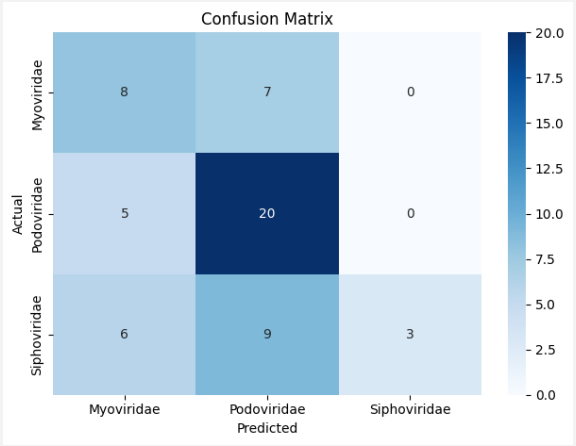


phage	Host	Protein Type	RBP Accession	Phage accession nr	Morphotype	CleanSeq	FAS
Bact	Escherichia	tailspike	YP_004782195.1	NC_015933.1	Podoviridae	MNEMFSQGG	>YP_004782195.1
Ba	Escherichia	tailspike	YP_009168860.1	NC_027994.1	Podoviridae	MSSGCGDVLS	>YP_009168
B	cherichia	tail spike	Q04830	NC_007636.1	Podoviridae	MSTITQFPSG	>Q04
	hia	tailspike	AEM91897.1	JN593240	Siphoviridae	MTRNVEELF	
		tailspike	AEM91898.1	JN593240	Siphoviridae	MISQ	
		spike	AEM91899.1	JN593240	Siphoviridae		
			YP_009880322.1	NC_049433.1	Siphovirid		
			NP_112089.1	NC_002730.1	Si		
			YP_512279.1	NC_007804.2			
			09572.1	FR775895.2			
			.1	NC_008			
Bacteri							
Bacteriophag							
Bacteriophage PhiEc							
Bacteriophage vB_EcoP_G							
Bacteriophage vB_EcoS_AKFV33							
Bacteriophage K1-5	Es						
Bacteriophage vB_CkoM_CkP1	Salmonel						
Bacteriophage T4	Entero						
Bacteriophage vB_SauM_LM12							
Bacteriophage KpnM6E1							
Bacteriophage T5							
Bacteriophage M							
Bacterioph							
Sal							
				NC_0			
			002f0c	NC_002747			
			pdb_00005e7f	NC_002703			
			pdb_0000q0pdk6	NC_004166			
			pdb_00008xci	NC_001416	Siphov		
		er	pdb_00005efv	NC_004615	Siphoviridae		
		ail fiber	pdb_00007qlr	CP011970.1	Siphoviridae	MS	
		putative structural	pdb_00004rga	NC_027120.1	Siphoviridae	GVLKGI	
	vobacteria	long tail fiber	P03744	NC_000866.4	Myoviridae	MATLKQIQFK	>P
Ba	Enterobacteria	short tail fiber	P10930	NC_000866.4	Myoviridae	MSNNTYQHV	>P10930
Bac	Escherichia	tail fiber	P0DJY6	AF083977.1	Myoviridae	MPKSTIIQNLC	>P0DJY6 tail fib

Validated RBPs after Processement with Python

Validated RBPs Analysis K-means and Confusion Matrix

```python
import numpy as np
from sklearn.cluster import KMeans
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.metrics import silhouette_score, classification_report
import matplotlib.pyplot as plt
import seaborn as sns

# === Load Data ===
df = pd.read_excel("validated_processed.xlsx")
sequences = df["CleanSeq"].astype(str).tolist()

# === Vectorize Sequences with TF-IDF 3-4-mers ===
vectorizer = TfidfVectorizer(analyzer='char', ngram_range=(3, 4))
X = vectorizer.fit_transform(sequences)

# === Find Optimal K Using Silhouette Score ===
best_k = None
best_score = -1
score_dict = {}

print("\nFinding optimal number of clusters...")
for k in range(2, 10):
    kmeans = KMeans(n_clusters=k, random_state=42, n_init='auto')
    labels = kmeans.fit_predict(X)
    score = silhouette_score(X, labels)
    score_dict[k] = score
    print(f"Silhouette score for k={k}: {score:.4f}")
    if score > best_score:
        best_score = score
        best_k = k

print(f"\nBest number of clusters: {best_k} (score={best_score:.4f})")

# === Final KMeans with Best K ===
kmeans = KMeans(n_clusters=best_k, random_state=42, n_init='auto')
df["BestKMeans_Cluster"] = kmeans.fit_predict(X)

# === Majority Vote Mapping from Cluster to Morphotype ===
def assign_morphotypes_by_majority(df, cluster_col, morph_col):
    mapping = {}
    for cluster in df[cluster_col].unique():
        subset = df[df[cluster_col] == cluster]
        most_common = subset[morph_col].mode()
        if not most_common.empty:
            mapping[cluster] = most_common.iloc[0]
        else:
            mapping[cluster] = "Unknown"
    return mapping

def apply_cluster_to_morphotype_mapping(df, cluster_col, mapping):
    return df[cluster_col].map(mapping).fillna("Unknown")

cluster_to_morph = assign_morphotypes_by_majority(df, "BestKMeans_Cluster", "Morphotype")
df["Predicted_morphotype"] = apply_cluster_to_morphotype_mapping(df, "BestKMeans_Cluster", cluster_to_morph)

print(f"\nCluster to Morphotype Mapping:\n{cluster_to_morph}")
# === Evaluate ===
valid_rows = df["Predicted_morphotype"].notna()
print("\nClassification Report:")
print(classification_report(df.loc[valid_rows, "Morphotype"], df.loc[valid_rows, "Predicted_morphotype"], zero_division=0))

# === Confusion Matrix ===
conf_mat = pd.crosstab(df["Morphotype"], df["Predicted_morphotype"], rownames=["Actual"], colnames=["Predicted"])
sns.heatmap(conf_mat, annot=True, cmap="Blues", fmt='d')
plt.title("Confusion Matrix")
plt.tight_layout()
plt.show()

# === Save Results ===
df.to_excel("phage_clustered_with_morphotypes.xlsx", index=False)
print("\nSaved results to 'phage_clustered_with_morphotypes.xlsx'")
```



Performance Analysis:

Podoviridae:

Highest recall and f1-score
↓
Reliably identified

Myoviridae and Siphoviridae:

Weaker performance;
Low recall for Siphoviridae
↓
Potential issues in classification

```
Classification Report:
              precision    recall  f1-score   support

  Myoviridae       0.42      0.53      0.47        15
 Podoviridae       0.56      0.80      0.66        25
Siphoviridae       1.00      0.17      0.29        18

    accuracy                           0.53        58
   macro avg       0.66      0.50      0.47        58
weighted avg       0.66      0.53      0.49        58
```

```
Morphotype
Podoviridae     25
Siphoviridae    18
Myoviridae      15
Name: count, dtype: int64
Predicted_morphotype
Podoviridae     36
Myoviridae      19
Siphoviridae     3
Name: count, dtype: int64
```

Validated RBPs – PCA and Clusters

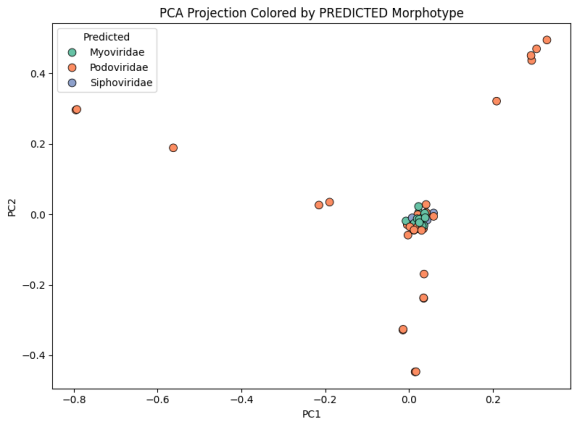
```python
# === PCA Visualization ===
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X.toarray())

# Add coordinates to dataframe
df["PC1"] = X_pca[:, 0]
df["PC2"] = X_pca[:, 1]

# === Plot PCA Results ===
fig, axes = plt.subplots(1, 2, figsize=(16, 6))  # 1 row, 2 columns

# Plot by True Morphotype
sns.scatterplot(data=df, x="PC1", y="PC2", hue="Morphotype", palette="Set2",
                s=60, edgecolor='k', ax=axes[0])
axes[0].set_title("PCA Projection Colored by TRUE Morphotype")
axes[0].legend(title="Morphotype", loc='best')

# Plot by Predicted Morphotype
sns.scatterplot(data=df, x="PC1", y="PC2", hue="Predicted_morphotype", palette="Set2",
                s=60, edgecolor='k', ax=axes[1])
axes[1].set_title("PCA Projection Colored by PREDICTED Morphotype")
axes[1].legend(title="Predicted", loc='best')

plt.tight_layout()
plt.show()
```

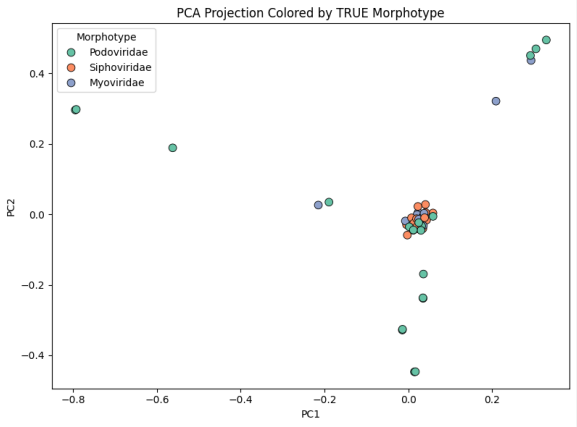
PCA - TRUE Morphotype Clusters:
Podoviridae: Well-separated
Siphoviridae: Cluster closer to Myoviridae
Myoviridae: Not clearly distinct; Slightly overlapping with Siphoviridae

- Distinct clustering for Podoviridae
- Confusion/blending between Siphoviridae and Myoviridae

PCA - PREDICTED Morphotype Clusters:
Podoviridae: New clustering observed, likely misclassified from the TRUE representation.
Siphoviridae: Predicted model seems less accurate
Myoviridae: Some appear clustered together with Podoviridae

- Predictive model not capture morphotype distinctions effectively



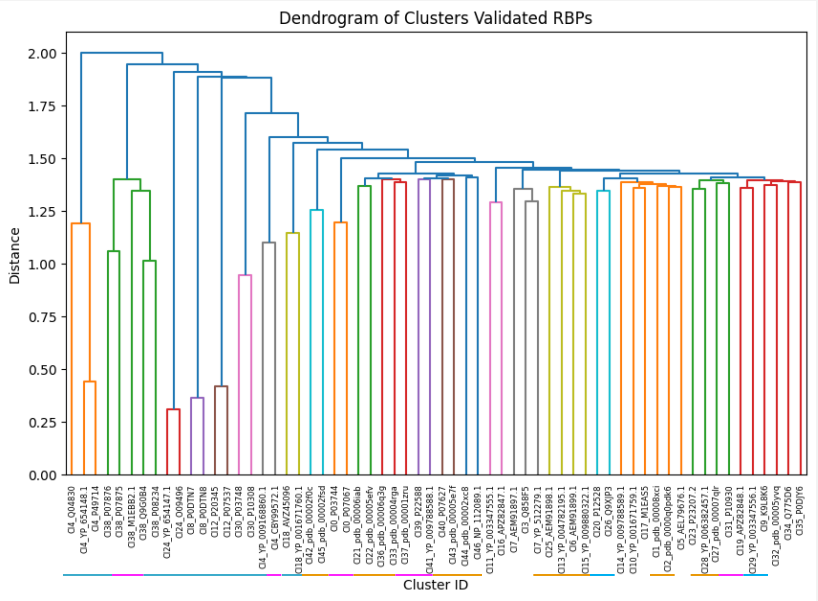


1st Dataset – Dendrogram K-means

```python
# === Dendrogram Visualization ===
# Compute the distance matrix
distance_matrix = pairwise_distances(X.toarray(), metric='cosine')

# Create the linkage matrix
Z = linkage(distance_matrix, method='ward')

# Plotting the dendrogram
plt.figure(figsize=(10, 6))
labels = df.apply(lambda row: f"{row['Morphotype']}_{row['RBP Accession']}", axis=1).values
dendrogram(Z, labels=labels)
plt.title('Dendrogram of Clusters Validated RBPs')
plt.xlabel('Cluster ID')
plt.ylabel('Distance')
plt.show()
```

For global compositional similarity - identifying potential functional groups:

- Cluster shape: Compact, evenly sized clusters
- Separation between clusters: Distinct groups shared sequence features
- Reflecting functional or structural similarity, even among non-homologous proteins - spacing on the left side indicate relationships among RBPs
- Clustering suggests distinct groups among the validated RBPs – some by morphotype



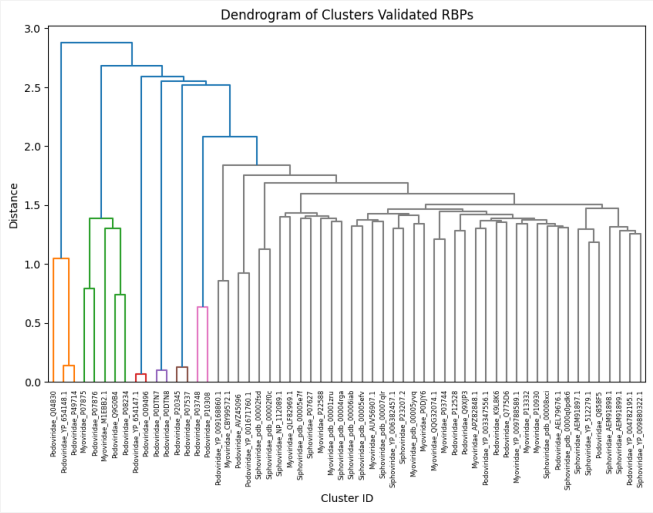
1st Dataset – Validated RBPs CD-HIT

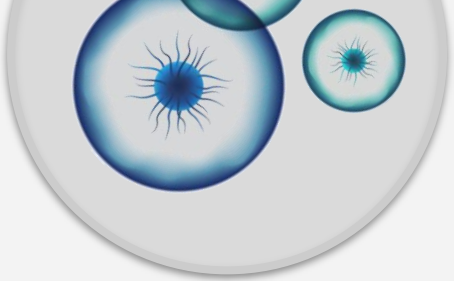
```python
from Bio import SeqIO
import pandas as pd
def parse_cdhit_output(file_path):
    clusters = {}
    current_cluster = None

    with open(file_path, 'r') as file:
        for line in file:
            line = line.strip()
            if line.startswith(">Cluster"):
                current_cluster = line.split(" ")[1]
                clusters[current_cluster] = []
            elif current_cluster is not None and line:
                if '>' in line:
                    try:
                        seq_info = line.split(",")[1] if ',' in line else line
                        seq_id_part = seq_info.split(">")[1]
                        seq_id = seq_id_part.split("...")[0].strip()
                        if seq_id:
                            clusters[current_cluster].append(seq_id)
                    except Exception as e:
                        print(f"Error parsing line: {line}\n{e}")
    return clusters

def load_fasta_sequences(fasta_path):
    seq_dict = {}
    for record in SeqIO.parse(fasta_path, "fasta"):
        core_id = record.id
        seq_dict[core_id] = str(record.seq)
    return seq_dict
def build_sequences_list(clusters, seq_dict):
    sequences = []
    data = []
    for cluster_id, seq_ids in clusters.items():
        for seq_id in seq_ids:
            seq = seq_dict.get(seq_id, "")
            if seq == "":
                print(f"Warning: Sequence ID '{seq_id}' not found in FASTA file.")
            data.append({'Cluster': cluster_id, 'Sequence ID': seq_id})
            sequences.append(seq)
    df = pd.DataFrame(data)
    return df, sequences

if __name__ == "__main__":
    cluster_file = "cluster_validadas.txt"
    fasta_file = "validated_RBP.fasta"

    clusters = parse_cdhit_output(cluster_file)
    seq_dict = load_fasta_sequences(fasta_file)
    df, sequences = build_sequences_list(clusters, seq_dict)

    print(f"Total sequences parsed: {len(sequences)}")

    df.to_csv("clustered_sequences0.csv", index=False)
```

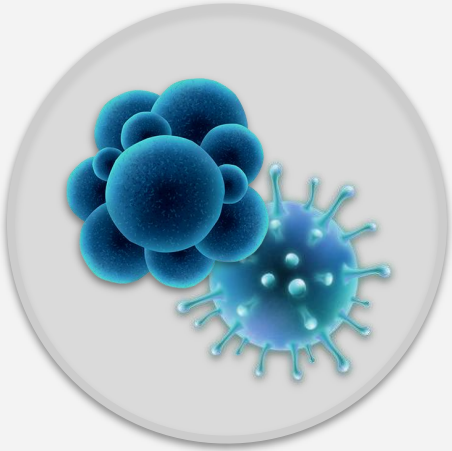
```python
from scipy.cluster.hierarchy import dendrogram, linkage
import numpy as np
from sklearn.feature_extraction.text import TfidfVectorizer

# Vectorize the sequences using TF-IDF
vectorizer = TfidfVectorizer(analyzer='char', ngram_range=(3, 4))
X = vectorizer.fit_transform(sequences)
# Create a linkage matrix using the actual feature matrix
Z = linkage(X.toarray(), method='ward')
# Plotting the dendrogram
plt.figure(figsize=(10, 6))

labels = df.apply(lambda row: f"Cl{row['Cluster']}_{row['Sequence ID']}", axis=1).values
dendrogram(Z, labels=labels)
plt.title('Dendrogram of Clusters Validated RBPs')
plt.xlabel('Cluster ID')
plt.ylabel('Distance')
plt.show()
```

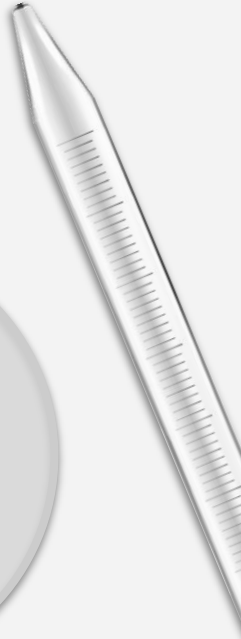
For redundancy removal and homology grouping:

- Pairwise sequence identity
- Cluster shape: Broader clusters
- Separation: Some branches merging at higher distances = low sequence similarity
- Bifurcations at grey represent a bigger diversification in less time comparing with the pink cluster with same ancestry
- Clustering suggests distinct groups among the validated RBPs – some by morphotype



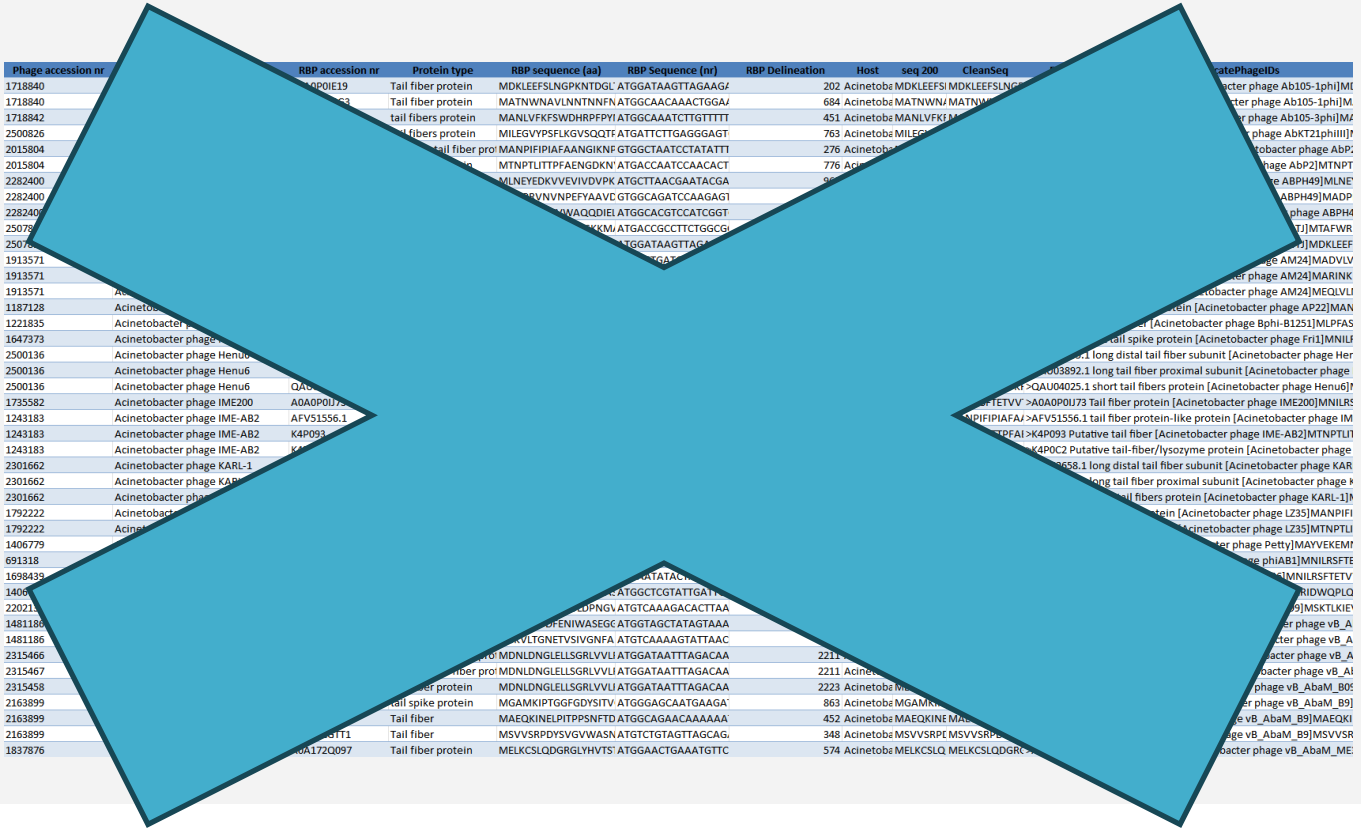






4.2

2nd Dataset



Phage accession nr		RBP accession nr	Protein type	RBP sequence (aa)	RBP Sequence (nr)	RBP Delineation	Host	seq 200	CleanSeq	CatePhageIDs
1718840			Tail fiber protein	MDKLEEFSLNGPKNTDGL	ATGGATAAGTTAGAAGA	202	Acinetoba	MDKLEEFSI	MDKLEEFSLNG	acter phage Ab105-1phi]MI
1718840			Tail fiber protein	MATNWNAVLNNTNNFN	ATGGCAACAAACTGGA	684	Acinetoba	MATNWN	MATNW	cter phage Ab105-1phi]M
1718842			tail fibers protein	MANLVFKFSWDHRPFPYI	ATGGCAAATCTTGTTTTT	451	Acinetoba	MANLVFKI	MA	r phage Ab105-3phi]MA
2500826			fibers protein	MILEGVYPSFLKGVSQQTF	ATGATTCTTGAGGGAGT	763	Acinetoba	MILEG	MILEG	r phage AbKT21phiIII]M
2015804			tail fiber pro	MANPIFIPIAFAANGIKNF	GTGGCTAATCCTATATTT	276	Acinetoba			tobacter phage AbP2
2015804				MTNPTLITTPFAENGDKN	ATGACCAATCCAACACT	776	Aci			hage AbP2]MTNPT
2282400				MLNEYEDKVVEVIVDVPK	ATGCTTAACGAATACGA	9				e ABPH49]MLNE
2282400				RVNVNPEFYAAVD	GTGGCAGATCCAAGAGT					ABPH49]MADP
2282401				QWAQQDIEI	ATGGCACGTCCATCGGT					phage ABPH4
2507[illegible]				KKM	ATGACCGCCTTCTGGCG					TJ]MTAFWR
2507[illegible]					TGGATAAGTTAGA					]MDKLEEF
1913571										ge AM24]MADVLV
1913571										er phage AM24]MARINK
1913571	A									tobacter phage AM24]MEQLVL
1187128	Acinetob									tein [Acinetobacter phage AP22]MAN
1221835	Acinetobacter p									r [Acinetobacter phage Bphi-B1251]MLPFAS
1647373	Acinetobacter phage									tail spike protein [Acinetobacter phage Fri1]MNILF
2500136	Acinetobacter phage Henu6									1 long distal tail fiber subunit [Acinetobacter phage Her
2500136	Acinetobacter phage Henu6									003892.1 long tail fiber proximal subunit [Acinetobacter phage
2500136	Acinetobacter phage Henu6	QA								>QAU04025.1 short tail fibers protein [Acinetobacter phage Henu6]M
1735582	Acinetobacter phage IME200	A0A0P0U7								TETVV >A0A0P0U73 Tail fiber protein [Acinetobacter phage IME200]MNILRS
1243183	Acinetobacter phage IME-AB2	AFV51556.1								NPIFIPIAFA >AFV51556.1 tail fiber protein-like protein [Acinetobacter phage IM
1243183	Acinetobacter phage IME-AB2	K4P093								TPFAI >K4P093 Putative tail fiber [Acinetobacter phage IME-AB2]MTNPTLI
1243183	Acinetobacter phage IME-AB2	K4								K4P0C2 Putative tail-fiber/lysozyme protein [Acinetobacter phage
2301662	Acinetobacter phage KARL-1									658.1 long distal tail fiber subunit [Acinetobacter phage KAR
2301662	Acinetobacter phage KAR									long tail fiber proximal subunit [Acinetobacter phage K
2301662	Acinetobacter pha									ail fibers protein [Acinetobacter phage KARL-1]M
1792222	Acinetobacte									otein [Acinetobacter phage LZ35]MANPIFI
1792222	Acinet									Acinetobacter phage LZ35]MTNPTLI
1406779										ter phage Petty]MAYVEKEM
691318										ge phiAB1]MNILRSFTE
1698439					GATATAC					MNILRSFTETV
1406					ATGGCTCGTATTGATT					RIDWQPLQ
220211				LDPNGV	ATGTCAAAGACACTTAA					9]MSKTLKIE
1481186				DFENIWASEGC	ATGGTAGCTATAGTAAA					er phage vB_A
1481186				RVLTGNETVSIVGNFA	ATGTCAAAAGTATTAAC					cter phage vB_A
2315466			pro	MDNLDNGLELLSGRLVVLI	ATGGATAATTTAGACAA	2211				acter phage vB_A
2315467			fiber pro	MDNLDNGLELLSGRLVVLI	ATGGATAATTTAGACAA	2211	Acinetob			bacter phage vB_AI
2315458			ber protein	MDNLDNGLELLSGRLVVLI	ATGGATAATTTAGACAA	2223	Acinetoba	M		phage vB_AbaM_B0
2163899			Tail spike protein	MGAMKIPTGGFGDYSITV	ATGGGAGCAATGAAGA	863	Acinetoba	MGAMK		er phage vB_AbaM_B9]
2163899			Tail fiber	MAEQKINELPITPPSNFTD	ATGGCAGAACAAAAAA	452	Acinetoba	MAEQKINE	MA	e vB_AbaM_B9]MAEQKI
2163899		STT1	Tail fiber	MSVVSRPDYSVGVWASN	ATGTCTGTAGTTAGCAG	348	Acinetoba	MSVVSRPI	MSVVSRPD	ge vB_AbaM_B9]MSVVSR
1837876		0A172Q097	Tail fiber protein	MELKCSLQDGRGLYHVTS	ATGGAACTGAAATGTTC	574	Acinetoba	MELKCSLQ	MELKCSLQDGRG	bacter phage vB_AbaM_ME

2nd Dataset – Computationally Predicted RBPs

```python
import pandas as pd
from Bio import Entrez, SeqIO
import time

# === Configuration ===
Entrez.email = "catiarosario10@gmail.com"
excel_path = "RBP_database_NaoValidadas.csv"
sheet_name = "RBP_database_NaoValidadas"
phage_id_col = "Bacteriophage"
protein_id_col = "RBP accession nr"
desc_col = "Protein type"
source_col = "Bacteriophage"
seq_col = "RBP sequence (aa)"
output_fasta_file = "unique_phage_proteins2.fasta"
output_excel_file = "phage_processed2.xlsx"

# === Helper Functions ===

def clean_sequence(seq):
    if not isinstance(seq, str):
        return ""
    lines = seq.strip().splitlines()
    lines = [line for line in lines if not line.startswith(">")]
    return "".join(lines).replace(" ", "").upper()

def make_fasta_header(row):
    desc = row.get(desc_col, "")
    source = row.get(source_col, "")
    desc = desc if pd.notna(desc) else ""
    source = source if pd.notna(source) else ""
    return f">{row[protein_id_col]} {desc} [{source}]".strip()

def make_full_fasta(row):
    header = make_fasta_header(row)
    seq = row["CleanSeq"]
    return f"{header}\n{seq}"
```

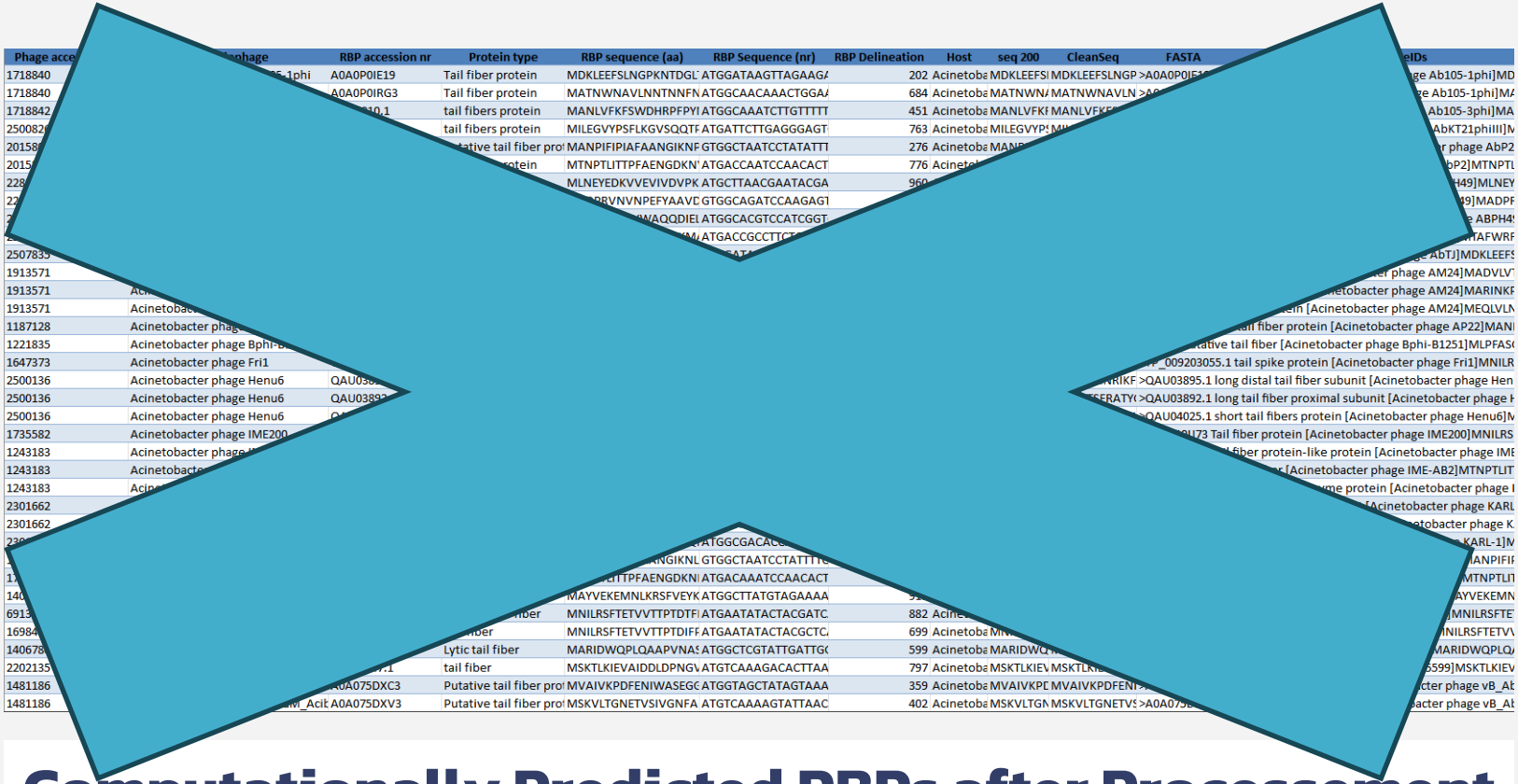
```python
def get_taxid_and_accession(phage_name):
    try:
        search = Entrez.esearch(db="nucleotide", term=phage_name, retmode="xml")
        record = Entrez.read(search)
        search.close()
        if not record["IdList"]:
            return None, None
        accession_id = record["IdList"][0]
        handle = Entrez.efetch(db="nucleotide", id=accession_id, rettype="gb", retmode="text")
        seq_record = SeqIO.read(handle, "genbank")
        handle.close()
        taxid = None
        for feature in seq_record.features:
            if feature.type == "source":
                for ref in feature.qualifiers.get("db_xref", []):
                    if ref.startswith("taxon:"):
                        taxid = ref.split(":")[1]
                        break
        return taxid, accession_id
    except Exception as e:
        print(f"[{phage_name}] TaxID/Accession fetch error: {e}")
        return None, None

def get_morphotype_from_taxonomy(taxid):
    try:
        handle = Entrez.efetch(db="taxonomy", id=taxid, retmode="xml")
        records = Entrez.read(handle)
        handle.close()
        lineage_ex = records[0].get("LineageEx", [])
        for entry in lineage_ex:
            if entry["Rank"] == "family":
                return entry["ScientificName"]
        return "Unknown morphotype"
    except Exception as e:
        print(f"[{taxid}] Taxonomy error: {e}")
        return "Error"
```

```python
# === Load data ===
df = pd.read_csv(excel_path)
df = df.dropna(subset=[phage_id_col, protein_id_col, seq_col])

# Step 1: Preserve original sequence
df["OriginalSeq"] = df[seq_col]

# Step 2: Clean sequences
df["CleanSeq"] = df[seq_col].apply(clean_sequence)

# Step 3: Create FASTA format
df["FASTA"] = df.apply(make_full_fasta, axis=1)

# Step 4: Add column with just the sequence part used in FASTA
df["SeqInFASTA"] = df["CleanSeq"]

# Step 5: Map duplicates
seq_to_phageids = {}
for _, row in df.iterrows():
    seq = row["CleanSeq"]
    phageid = row[phage_id_col]
    seq_to_phageids.setdefault(seq, []).append(phageid)

def get_duplicate_phageids(row):
    seq = row["CleanSeq"]
    phageid = row[phage_id_col]
    duplicates = [pid for pid in seq_to_phageids.get(seq, []) if pid != phageid]
    return ", ".join(duplicates) if duplicates else ""

df["DuplicatePhageIDs"] = df.apply(get_duplicate_phageids, axis=1)

# Step 6: Remove duplicate sequences, keep first occurrence
df_unique = df.drop_duplicates(subset=["CleanSeq"]).copy()

# Step 7: Retrieve morphotype and accession
taxid_map = {}
accession_map = {}
morphotype_map = {}

for name in df_unique[phage_id_col].astype(str).unique():
    print(f"Fetching info for: {name}")
    taxid, accession = get_taxid_and_accession(name)
    morphotype = get_morphotype_from_taxonomy(taxid) if taxid else "TaxID not found"
    taxid_map[name] = taxid
    accession_map[name] = accession
    morphotype_map[name] = morphotype
    time.sleep(0.4)

df_unique["TaxID"] = df_unique[phage_id_col].map(taxid_map)
df_unique["AccessionFetched"] = df_unique[phage_id_col].map(accession_map)
df_unique["Morphotype"] = df_unique[phage_id_col].map(morphotype_map)
# Step 8: Write FASTA file
with open(output_fasta_file, "w") as fasta_out:
    for fasta_entry in df_unique["FASTA"]:
        fasta_out.write(fasta_entry + "\n")

# Step 9: Save to Excel
df_unique.to_excel(output_excel_file, index=False)

print(f"FASTA saved to '{output_fasta_file}'")
print(f"Excel saved to '{output_excel_file}'")
```

Workflow data processing – NON-VALIDATED:

- Fetch the id of bacteriophages from name
- Add column with ids
- Add FASTA format if missing
- Create column with only aa to remove duplicates
- Create column with FASTA format
- Create new column with duplicated identifiers
- Remove duplicate lines based on sequence
- Retrieve morphotypes
- Create FASTA file and Excel file

2nd Dataset Processement with Python



Phage acce	phage	RBP accession nr	Protein type	RBP sequence (aa)	RBP Sequence (nr)	RBP Delineation	Host	seq 200	CleanSeq	FASTA	elDs
1718840		A0A0P0IE19	Tail fiber protein	MDKLEEFSLNGPKNTDGL	ATGGATAAGTTAGAAGA	202	Acinetoba	MDKLEEFS	MDKLEEFSLNGP	>A0A0P0IE	ge Ab105-1phi]MD
1718840		A0A0P0IRG3	Tail fiber protein	MATNWNAVLNNTNNFN	ATGGCAACAAACTGGAA	684	Acinetoba	MATNWN	MATNWNAVLN	>A0	e Ab105-1phi]MA
1718842		10.1	tail fibers protein	MANLVFKFSWDHRPFPYI	ATGGCAAATCTTGTTTTT	451	Acinetoba	MANLVFKF	MANLVFKF		Ab105-3phi]MA
2500820			tail fibers protein	MILEGVYPSFLKGVSQQTF	ATGATTCTTGAGGGAGT	763	Acinetoba	MILEGVYPS	MI		AbKT21phIII]M
2015			tative tail fiber pro	MANPIFIPIAFAANGIKNF	GTGGCTAATCCTATATTT	276	Acinetoba	MANP			r phage AbP2
2015			rotein	MTNPTLITTPFAENGDKN	ATGACCAATCCAACACT	776	Acinetoba				bP2]MTNPTL
228				MLNEYEDKVVEVIVDVPK	ATGCTTAACGAATACGA	960					49]MLNEY
22				PRVNVNPEFYAAVD	GTGGCAGATCCAAGAG						49]MADPF
2				WAQQDIEL	ATGGCACGTCCATCGGT						e ABPH4
2				VM	ATGACCGCCTTCT						TAFWRF
2507835											e AbTJ]MDKLEEF
1913571											r phage AM24]MADVL
1913571	Ac										etobacter phage AM24]MARINKF
1913571	Acinetobac										in [Acinetobacter phage AM24]MEQLVL
1187128	Acinetobacter phag										tail fiber protein [Acinetobacter phage AP22]MAN
1221835	Acinetobacter phage Bphi-B										tative tail fiber [Acinetobacter phage Bphi-B1251]MLPFAS
1647373	Acinetobacter phage Fri1										_009203055.1 tail spike protein [Acinetobacter phage Fri1]MNILR
2500136	Acinetobacter phage Henu6	QAU038							NRIKF	>QAU03895.1 long distal tail fiber subunit [Acinetobacter phage Hen	
2500136	Acinetobacter phage Henu6	QAU0389							SERATY(	>QAU03892.1 long tail fiber proximal subunit [Acinetobacter phage H	
2500136	Acinetobacter phage Henu6	QA								QAU04025.1 short tail fibers protein [Acinetobacter phage Henu6]M	
1735582	Acinetobacter phage IME200									73 Tail fiber protein [Acinetobacter phage IME200]MNILRS	
1243183	Acinetobacter phage									fiber protein-like protein [Acinetobacter phage IME	
1243183	Acinetobacter									[Acinetobacter phage IME-AB2]MTNPTLIT	
1243183	Acin									me protein [Acinetobacter phage I	
2301662										[Acinetobacter phage KARL	
2301662										otobacter phage K	
230				ATGGCGACAC						KARL-1]M	
				NGIKNL	GTGGCTAATCCTATTTT						ANPIFIP
17				LITTPFAENGDKN	ATGACAAATCCAACACT						MTNPTLI
140				MAYVEKEMNLKRSFVEYK	ATGGCTTATGTAGAAAA	9					AYVEKEMN
6913			mber	MNILRSFTETVVTPTDTF	ATGAATATACTACGATC	882	Acine				MNILRSFTE
16984			mber	MNILRSFTETVVTPTDIFF	ATGAATATACTACGCTC	699	Acinetoba	MN			NILRSFTETV
140678			Lytic tail fiber	MARIDWQPLQAAPVNAS	ATGGCTCGTATTGATTG	599	Acinetoba	MARIDWQ			MARIDWQPLQ
2202135		7.1	tail fiber	MSKTLKIEVAIDDLDPNG	ATGTCAAAGACACTTAA	797	Acinetoba	MSKTLKIE	MSKTLKIE		6599]MSKTLKIEV
1481186		A0A075DXC3	Putative tail fiber pro	MVAIVKPDFENIWASEGG	ATGGTAGCTATAGTAAA	359	Acinetoba	MVAIVKPD	MVAIVKPDFEN		cter phage vB_Ab
1481186	M_Acit	A0A075DXV3	Putative tail fiber pro	MSKVLTGNETVSIVGNFA	ATGTCAAAAGTATTAAC	402	Acinetoba	MSKVLTGN	MSKVLTGNETV	>A0A075D	acter phage vB_Ab

Computationally Predicted RBPs after Processement with Python

2nd Dataset Analysis – K-means and PCA

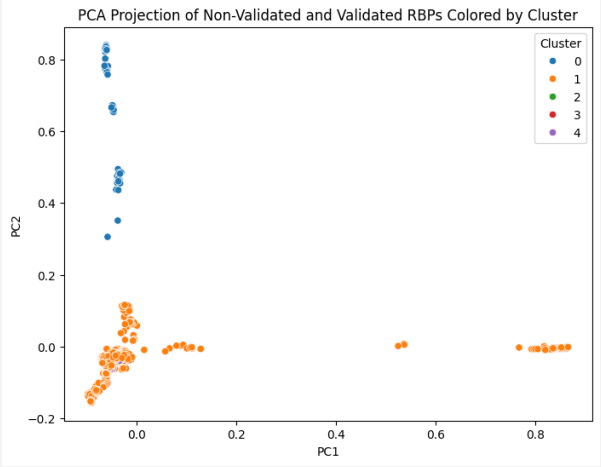
```python
import pandas as pd
from sklearn.cluster import KMeans
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.metrics import silhouette_score
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.decomposition import PCA
from sklearn.metrics import pairwise_distances
from scipy.cluster.hierarchy import dendrogram, linkage

# === Load data ===
df = pd.read_excel("phage_processed2.xlsx")
sequences = df["RBP sequence (aa)"].astype(str).tolist()

# === Vectorize sequences using TF-IDF of 3-4-mers ===
vectorizer = TfidfVectorizer(analyzer='char', ngram_range=(3, 4))
X = vectorizer.fit_transform(sequences)

# === Best number of clusters by Silhouette score ===
print("Finding optimal number of clusters by silhouette score...")

best_k = None
best_score = -1
scores = []

for k in range(2, 6):
    kmeans = KMeans(n_clusters=k, random_state=42, n_init='auto')
    labels = kmeans.fit_predict(X)
    score = silhouette_score(X, labels)
    scores.append(score)
    print(f"k={k}, silhouette score={score:.4f}")
    if score > best_score:
        best_score = score
        best_k = k

print(f"\nBest k={best_k} with silhouette score={best_score:.4f}")

kmeans = KMeans(n_clusters=best_k, random_state=42, n_init='auto')
df['Cluster'] = kmeans.fit_predict(X)

print(f"Best number of clusters: {best_k} with silhouette score: {best_score:.4f}")

best_model = KMeans(n_clusters=best_k, random_state=42, n_init='auto')
df['Cluster'] = best_model.fit_predict(X)
```

```
Finding optimal number of clusters by silhouette score...
k=2, silhouette score=0.0163
k=3, silhouette score=0.0252
k=4, silhouette score=0.0324
k=5, silhouette score=0.0393

Best k=5 with silhouette score=0.0393
Best number of clusters: 5 with silhouette score: 0.0393
```

```python
# === PCA ===
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X.toarray())
df['PC1'] = X_pca[:, 0]
df['PC2'] = X_pca[:, 1]

plt.figure(figsize=(8,6))
sns.scatterplot(data=df, x='PC1', y='PC2', hue='Cluster', palette='tab10', legend='full')
plt.title('PCA Projection of Non-Validated RBPs Colored by Cluster')
plt.show()

# === Dendrogram Visualization ===
distance_matrix = pairwise_distances(X.toarray(), metric='cosine')

Z = linkage(distance_matrix, method='ward')

plt.figure(figsize=(30, 10))
labels = df.apply(lambda row: f"Cl{row['Cluster']}_{row['RBP accession nr']}", axis=1).values
dendrogram(Z, labels=labels, truncate_mode='level', p=10, leaf_rotation=90)
plt.xticks(rotation=90, size=8)
plt.title('Dendrogram of Non-Validated RBPs Clustering')
plt.xlabel('Sample Index or Cluster ID')
plt.ylabel('Distance')
plt.show()

# === Save results ===
df.to_excel("rbp_clustered_results_onlyNV.xlsx", index=False)
print("Saved clustering results to rbp_clustered_results_onlyNV.xlsx")
```

```python
best_score = -1
best_params = {}

for k in range(2, 10):
    for init_method in ['k-means++', 'random']:
        model = KMeans(n_clusters=k, init=init_method, random_state=42, n_init='auto')
        labels = model.fit_predict(X)
        score = silhouette_score(X, labels)
        if score > best_score:
            best_score = score
            best_params = {'n_clusters': k, 'init': init_method}

print(f"Best params: {best_params} with silhouette score: {best_score:.4f}")
```

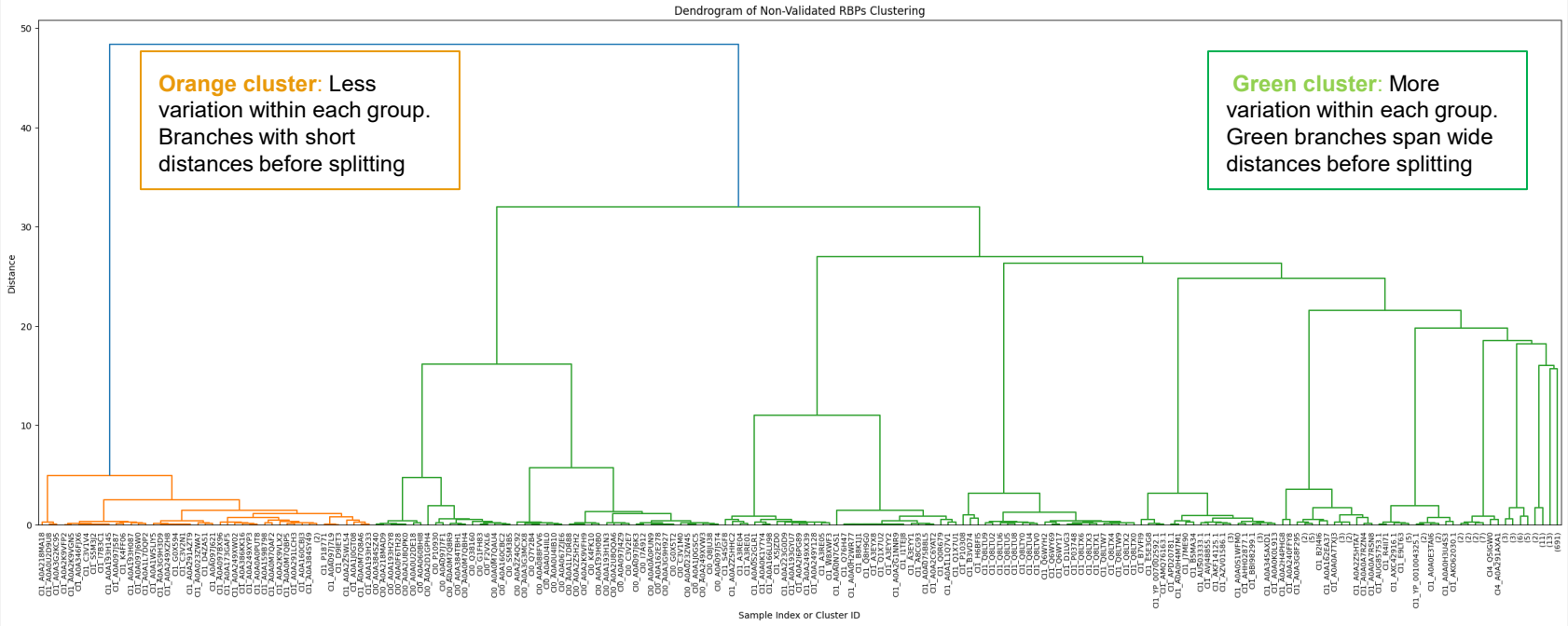
```
Best params: {'n_clusters': 9, 'init': 'random'} with silhouette score: 0.0810
```



Cluster 0:
Separated from others → distinct characteristics
Cluster 1:
Larger group with some spread along both axes
Cluster 2, Cluster 3, Cluster 4:
Less populated and overlap → similarity in characteristics

Interpretation:
- Distinct clusters → Non-Validated RBPs may have different characteristics
- **Cluster 0** is notably distinct, potentially representing a unique group of RBPs

2nd Dataset Analysis – Dendrogram



- Cluster shape: Broad
- Separation between clusters: High
- Cohesion between sub-groups
- Nº Clusters: Low

- Dissimilar sequences that share functional elements

These clusters likely represent:

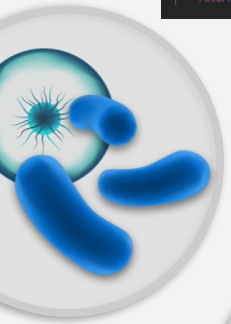
- Possibly novel or divergent RBPs
- Groupings of functionally analogous

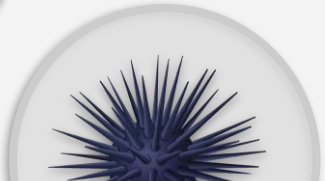
2nd Dataset – Non-Validated RBPs CD-HIT

```python
from Bio import SeqIO
import pandas as pd
def parse_cdhit_output(file_path):
    clusters = {}
    current_cluster = None

    with open(file_path, 'r') as file:
        for line in file:
            line = line.strip()
            if line.startswith(">Cluster"):
                current_cluster = line.split(" ")[1]
                clusters[current_cluster] = []
            elif current_cluster is not None and line:
                if '>' in line:
                    try:
                        seq_info = line.split(",")[1] if ',' in line else line
                        seq_id_part = seq_info.split(">")[1]
                        seq_id = seq_id_part.split("...")[0].strip()
                        if seq_id:
                            clusters[current_cluster].append(seq_id)
                    except Exception as e:
                        print(f"Error parsing line: {line}\n{e}")
    return clusters

def load_fasta_sequences(fasta_path):
    seq_dict = {}
    for record in SeqIO.parse(fasta_path, "fasta"):
        core_id = record.id
        seq_dict[core_id] = str(record.seq)
    return seq_dict
```

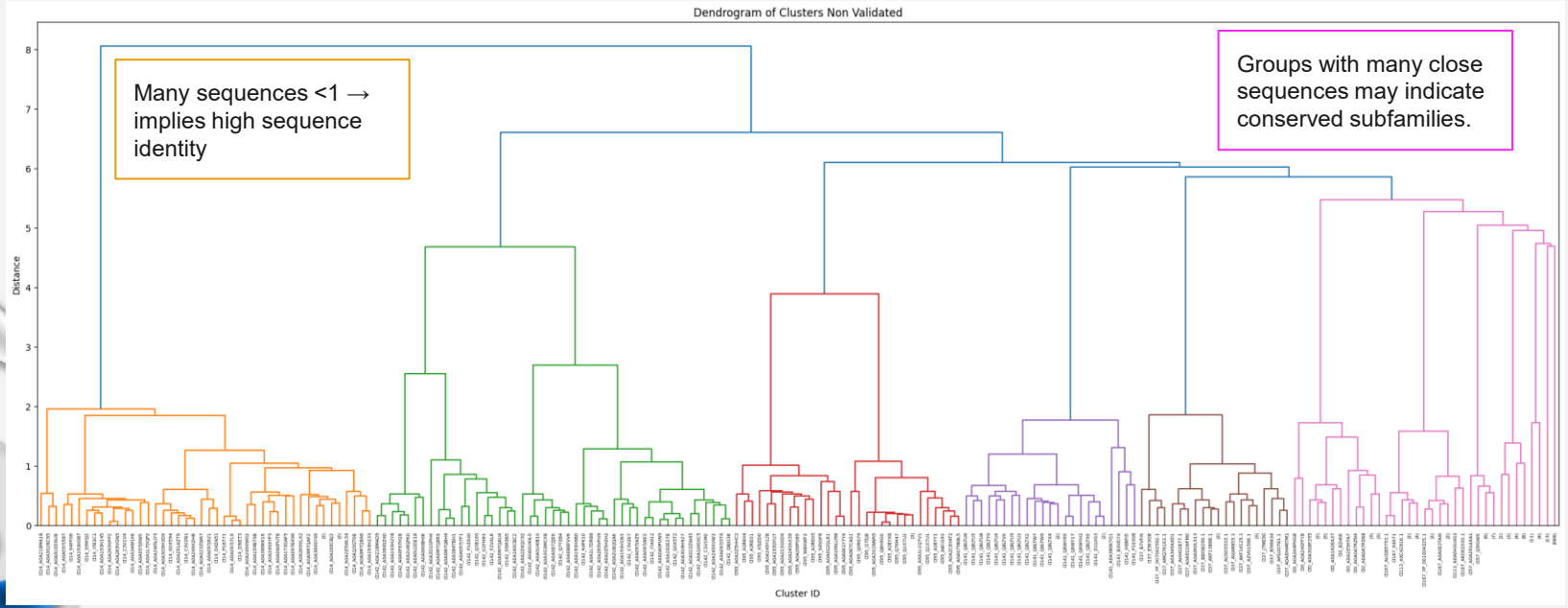
```python
def build_sequences_list(clusters, seq_dict):
    sequences = []
    data = []
    for cluster_id, seq_ids in clusters.items():
        for seq_id in seq_ids:
            seq = seq_dict.get(seq_id, "")
            if seq == "":
                print(f"Warning: Sequence ID '{seq_id}' not found in FASTA file.")
            data.append({'Cluster': cluster_id, 'Sequence ID': seq_id})
            sequences.append(seq)
    df = pd.DataFrame(data)
    return df, sequences

def build_sequences_list(clusters, seq_dict):
    sequences = []
    data = []
    for cluster_id, seq_ids in clusters.items():
        for seq_id in seq_ids:
            seq = seq_dict.get(seq_id, "")
            if seq == "":
                print(f"Warning: Sequence ID '{seq_id}' not found in FASTA file.")
            data.append({'Cluster': cluster_id, 'Sequence ID': seq_id})
            sequences.append(seq)
    df = pd.DataFrame(data)
    return df, sequences

if __name__ == "__main__":
    cluster_file = "cluster_naovalidadas.txt"
    fasta_file = "unique_phage_proteins2.fasta"

    clusters = parse_cdhit_output(cluster_file)
    seq_dict = load_fasta_sequences(fasta_file)
    df, sequences = build_sequences_list(clusters, seq_dict)

    print(f"Total sequences parsed: {len(sequences)}")
    df.to_csv("clustered_sequences2.csv", index=False)
```

```python
from scipy.cluster.hierarchy import dendrogram, linkage
import numpy as np
from sklearn.feature_extraction.text import TfidfVectorizer

# Vectorize the sequences using TF-IDF
vectorizer = TfidfVectorizer(analyzer='char', ngram_range=(3, 4))
X = vectorizer.fit_transform(sequences)  # This is your feature matrix
# Create a linkage matrix using the actual feature matrix
Z = linkage(X.toarray(), method='ward')  # Convert sparse matrix to dense

plt.figure(figsize=(30, 10))
labels = df.apply(lambda row: f"Cl row['Cluster']}_{row['Sequence ID']}", axis=1).values
dendrogram(Z, labels=labels, truncate_mode='level', p=10, leaf_rotation=90)
plt.xticks(rotation=90, size=5)
plt.title('Dendrogram of Clusters Non Validated')
plt.xlabel('Cluster ID')
plt.ylabel('Distance')
plt.show()
```





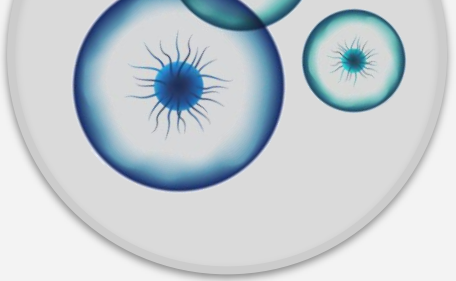
2nd Dataset – Non-Validated RBPs CD-HIT

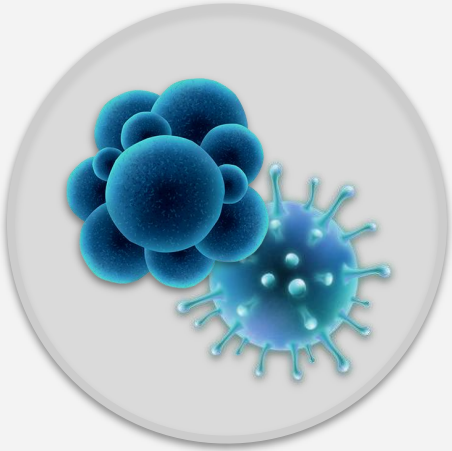


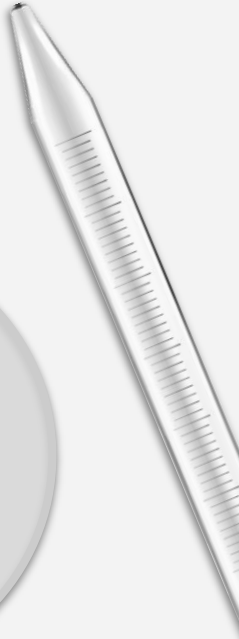
- Cluster shape: Compact, narrow (short branches within)
- Separation between clusters: High (clear distinction)
- Intra-cluster similarity: High (identity ≥ 40%)
- Inter-cluster difference: High (hard cutoff)

Each cluster could correspond to:

- One phage RBP subtype,
- A known structural domain or tail fiber class,
- Evolutionary lineages with strong sequence conservation.

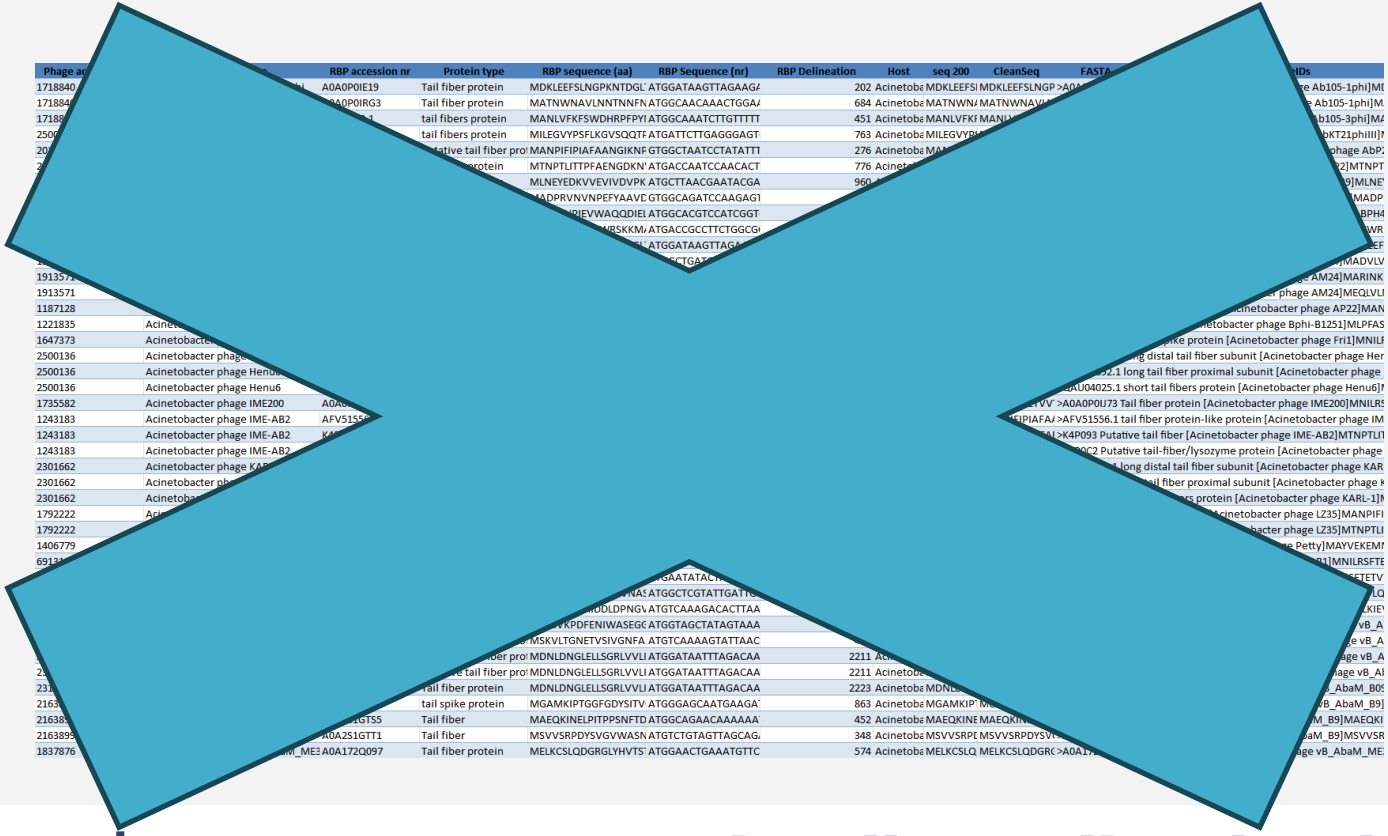






4.3

3rd Dataset



Phage ac		RBP accession nr	Protein type	RBP sequence (aa)	RBP Sequence (nr)	RBP Delineation	Host	seq 200	CleanSeq	FASTA	IDs
1718840		A0A0P0IE19	Tail fiber protein	MDKLEEFSLNGPKNTDGL	ATGGATAAGTTAGAAGA	202	Acinetoba	MDKLEEFSI	MDKLEEFSLNGP	>A0A	e Ab105-1phi]MI
171884		0A0P0IRG3	Tail fiber protein	MATNWNAVLNNTNNFN	ATGGCAACAAACTGGA	684	Acinetoba	MATNWNA	MATNWNAVL		e Ab105-1phi]M
17188		1	tail fibers protein	MANLVFKFSWDHRPFPYI	ATGGCAAATCTTGTTTTT	451	Acinetoba	MANLVFKI	MANL		b105-3phi]MA
2500			tail fibers protein	MILEGVYPSFLKGVSQQTF	ATGATTCTTGAGGGAGT	763	Acinetoba	MILEGVYP			bKT21phiIII]M
20			tative tail fiber pro	MANPIFIPIAFAANGIKNF	GTGGCTAATCCTATATTT	276	Acinetoba	MAN			phage AbP2
2			protein	MTNPTLITTPFAENGDKN	ATGACCAATCCAACACT	776	Acinetoba				P2]MTNPT
				MLNEYEDKVVEVIVDVPK	ATGCTTAACGAATACGA	960					9]MLNE
				MADPRVNVNPEFYAAVD	GTGGCAGATCCAAGAGT						MADP
				VIEVWAQQDIEL	ATGGCACGTCCATCGGT						BPH4
				VRSKKM	ATGACCGCCTTCTGGCG						WR
					ATGGATAAGTTAGA						EF
											MADVLV
191357											AM24]MARINK
1913571											phage AM24]MEQLVLI
1187128											cinetobacter phage AP22]MAN
1221835	Acinet										netobacter phage Bphi-B1251]MLPFAS
1647373	Acinetobacte										like protein [Acinetobacter phage Fri1]MNILI
2500136	Acinetobacter phage										ng distal tail fiber subunit [Acinetobacter phage Her
2500136	Acinetobacter phage Henu										52.1 long tail fiber proximal subunit [Acinetobacter phage
2500136	Acinetobacter phage Henu6										AU04025.1 short tail fibers protein [Acinetobacter phage Henu6]M
1735582	Acinetobacter phage IME200	A0A									VV >A0A0P0IJ73 Tail fiber protein [Acinetobacter phage IME200]MNILRS
1243183	Acinetobacter phage IME-AB2	AFV5155									HPIAFAA >AFV51556.1 tail fiber protein-like protein [Acinetobacter phage IM
1243183	Acinetobacter phage IME-AB2	K4									1>K4P093 Putative tail fiber [Acinetobacter phage IME-AB2]MTNPTLI
1243183	Acinetobacter phage IME-AB2										0C2 Putative tail-fiber/lysozyme protein [Acinetobacter phage
2301662	Acinetobacter phage KAR										long distal tail fiber subunit [Acinetobacter phage KAR
2301662	Acinetobacter ph										ail fiber proximal subunit [Acinetobacter phage K
2301662	Acinetoba										rs protein [Acinetobacter phage KARL-1]M
1792222	Aci										Acinetobacter phage LZ35]MANPIFI
1792222											bacter phage LZ35]MTNPTLI
1406779											e Petty]MAYVEKEM
691					GAATATACH						1]MNILRSFTE
											5TETV
				NA	ATGGCTCGTATTGATTG						LQ
				MDDLDPNGV	ATGTCAAAGACACTTAA						LKIE
				VKPDFENIWASEGC	ATGGTAGCTATAGTAAA						vB_A
				MSKVLTGNETVSIVGNFA	ATGTCAAAAGTATTAAC						e vB_A
			iber pro	MDNLDNGLELLSGRLVVLI	ATGGATAATTTAGACAA	2211	Ac				age vB_A
2			e tail fiber pro	MDNLDNGLELLSGRLVVLI	ATGGATAATTTAGACAA	2211	Acinetoba				hage vB_A
231			Tail fiber protein	MDNLDNGLELLSGRLVVLI	ATGGATAATTTAGACAA	2223	Acinetoba	MDNL			_AbaM_B0
2163			tail spike protein	MGAMKIPTGGFGDYSITV	ATGGGAGCAATGAAGA	863	Acinetoba	MGAMKIP	M		B_AbaM_B9]
21638		1GTS5	Tail fiber	MAEQKINELPITPPSNFTD	ATGGCAGAACAAAAAA	452	Acinetoba	MAEQKIN	MAEQKIN		M_B9]MAEQKI
216389		A0A2S1GTT1	Tail fiber	MSVVSRPDYSVGVWASN	ATGTCTGTAGTTAGCAG	348	Acinetoba	MSVVSRP	MSVVSRPDYSV		baM_B9]MSVVSR
1837876	M_ME	A0A172Q097	Tail fiber protein	MELKCSLQDGRGLYHVTS	ATGGAACTGAAATGTTC	574	Acinetoba	MELKCSLQ	MELKCSLQDGR	>A0A1	ge vB_AbaM_ME

3nd Dataset – Computationally Predicted and Validated RBPs

```python
import pandas as pd
from Bio import Entrez, SeqIO
import time

# === Configuration ===
Entrez.email = "catiarosario10@gmail.com"
excel_path = "data_test_code.xlsx"
sheet_name = "data_test"
phage_id_col = "Bacteriophage"
protein_id_col = "RBP_Accession"
desc_col = "Protein_Type"
source_col = "Bacteriophage"
seq_col = "RBP_Sequence"
output_fasta_file = "unique_phage_proteins1.fasta"
output_excel_file = "phage_processed1.xlsx"

# === Helper Functions ===

def clean_sequence(seq):
    if not isinstance(seq, str):
        return ""
    lines = seq.strip().splitlines()
    lines = [line for line in lines if not line.startswith(">")]
    return "".join(lines).replace(" ", "").upper()

def make_fasta_header(row):
    desc = row.get(desc_col, "")
    source = row.get(source_col, "")
    desc = desc if pd.notna(desc) else ""
    source = source if pd.notna(source) else ""
    return f">{row[protein_id_col]} {desc} [{source}]".strip()

def make_full_fasta(row):
    header = make_fasta_header(row)
    seq = row["CleanSeq"]
    return f"{header}\n{seq}"
```

```python
def get_taxid_and_accession(phage_name):
    try:
        search = Entrez.esearch(db="nucleotide", term=phage_name, retmode="xml")
        record = Entrez.read(search)
        search.close()
        if not record["IdList"]:
            return None, None
        accession_id = record["IdList"][0]
        handle = Entrez.efetch(db="nucleotide", id=accession_id, rettype="gb", retmode="text")
        seq_record = SeqIO.read(handle, "genbank")
        handle.close()
        taxid = None
        for feature in seq_record.features:
            if feature.type == "source":
                for ref in feature.qualifiers.get("db_xref", []):
                    if ref.startswith("taxon:"):
                        taxid = ref.split(":")[1]
                        break
        return taxid, accession_id
    except Exception as e:
        print(f"[{phage_name}] TaxID/Accession fetch error: {e}")
        return None, None

def get_morphotype_from_taxonomy(taxid):
    try:
        handle = Entrez.efetch(db="taxonomy", id=taxid, retmode="xml")
        records = Entrez.read(handle)
        handle.close()
        lineage_ex = records[0].get("LineageEx", [])
        for entry in lineage_ex:
            if entry["Rank"] == "family":
                return entry["ScientificName"]
        return "Unknown morphotype"
    except Exception as e:
        print(f"[{taxid}] Taxonomy error: {e}")
        return "Error"
```

```python
# === Load data ===
df = pd.read_excel(excel_path, sheet_name=sheet_name)
df = df.dropna(subset=[phage_id_col, protein_id_col, seq_col])

# Step 1: Preserve original sequence
df["OriginalSeq"] = df[seq_col]

# Step 2: Clean sequences
df["CleanSeq"] = df[seq_col].apply(clean_sequence)

# Step 3: Create FASTA format
df["FASTA"] = df.apply(make_full_fasta, axis=1)

# Step 4: Add column with just the sequence part used in FASTA
df["SeqInFASTA"] = df["CleanSeq"]

# Step 5: Map duplicates
seq_to_phageids = {}
for _, row in df.iterrows():
    seq = row["CleanSeq"]
    phageid = row[phage_id_col]
    seq_to_phageids.setdefault(seq, []).append(phageid)

def get_duplicate_phageids(row):
    seq = row["CleanSeq"]
    phageid = row[phage_id_col]
    duplicates = [pid for pid in seq_to_phageids.get(seq, []) if pid != phageid]
    return ", ".join(duplicates) if duplicates else ""

df["DuplicatePhageIDs"] = df.apply(get_duplicate_phageids, axis=1)

# Step 6: Remove duplicate sequences, keep first occurrence
df_unique = df.drop_duplicates(subset=["CleanSeq"]).copy()

# Step 7: Retrieve morphotype and accession
taxid_map = {}
accession_map = {}
morphotype_map = {}

for name in df_unique[phage_id_col].astype(str).unique():
    print(f"Fetching info for: {name}")
    taxid, accession = get_taxid_and_accession(name)
    morphotype = get_morphotype_from_taxonomy(taxid) if taxid else "TaxID not found"
    taxid_map[name] = taxid
    accession_map[name] = accession
    morphotype_map[name] = morphotype
    time.sleep(0.4)

df_unique["TaxID"] = df_unique[phage_id_col].map(taxid_map)
df_unique["AccessionFetched"] = df_unique[phage_id_col].map(accession_map)
df_unique["Morphotype"] = df_unique[phage_id_col].map(morphotype_map)

# Step 8: Write FASTA file
with open(output_fasta_file, "w") as fasta_out:
    for fasta_entry in df_unique["FASTA"]:
        fasta_out.write(fasta_entry + "\n")
# Step 9: Save to Excel
df_unique.to_excel(output_excel_file, index=False)

print(f"FASTA saved to '{output_fasta_file}'")
print(f"Excel saved to '{output_excel_file}'")
```

Workflow data processing – NON-VALIDATED:

- Fetch the id of bacteriophages from name
- Add column with ids
- Add FASTA format if missing
- Create column with only aa to remove duplicates
- Create column with FASTA format
- Create new column with duplicated identifiers
- Remove duplicate lines based on sequence
- Retrieve morphotypes
- Create FASTA file and Excel file

3rd Dataset Processement with Python

Phage ac	riophage	RBP accession nr	Protein type	RBP sequence (aa)	RBP Sequence (nr)	RBP Delineation	Host	seq 200	CleanSeq	FASTA	ageIDs
1718840	105-1phi	A0A0P0IE19	Tail fiber protein	MDKLEEFSLNGPKNTDGL	ATGGATAAGTTAGAAGA	202	Acinetoba	MDKLEEFSI	MDKLEEFSLNGP	>A0A0P0	phage Ab105-1phi]MD
1718840		A0A0P0IRG3	Tail fiber protein	MATNWNAVLNNTNNFN	ATGGCAACAAACTGGAA	684	Acinetoba	MATNWN/	MATNWNAVLN		hage Ab105-1phi]MA
1718841		9010.1	tail fibers protein	MANLVFKFSWDHRPFPYI	ATGGCAAATCTTGTTTTT	451	Acinetoba	MANLVFK	MANLVF		ge Ab105-3phi]MA
25008			tail fibers protein	MILEGVYPSFLKGVSQQTF	ATGATTCTTGAGGGAGT	763	Acinetoba	MILEGVYP			ge AbKT21phIII]M
2015			Putative tail fiber pro	MANPIFIPIAFAANGIKNF	GTGGCTAATCCTATATTT	276	Acinetoba	MA			cter phage AbP2
207			protein	MTNPTLITTPFAENGDKN	ATGACCAATCCAACACT	776	Acine				e AbP2]MTNPTL
2				MLNEYEDKVVEVIVDVPK	ATGCTTAACGAATACGA	9					BPH49]MLNEY
				ADPRVNVNPEFYAAVD	GTGGCAGATCCAAGAG	1					PH49]MADPF
				EVWAQQDIEL	ATGGCACGTCCATCGGT						age ABPH4
				KKM	ATGACCGCCTT						MTAFWRF
2507											ge AbTJ]MDKLEEFS
1913571											cter phage AM24]MADVLV
1913571											cinetobacter phage AM24]MARINKF
1913571	Acinetob										otein [Acinetobacter phage AM24]MEQLVLN
1187128	Acinetobacter ph										e tail fiber protein [Acinetobacter phage AP22]MAN
1221835	Acinetobacter phage Bph										Putative tail fiber [Acinetobacter phage Bphi-B1251]MLPFAS
1647373	Acinetobacter phage Fri1										YP_009203055.1 tail spike protein [Acinetobacter phage Fri1]MNILR
2500136	Acinetobacter phage Henu6	QAU0							NINRIKF	>QAU03895.1 long distal tail fiber subunit [Acinetobacter phage Hen	
2500136	Acinetobacter phage Henu6	QAU03							TSFRATY	>QAU03892.1 long tail fiber proximal subunit [Acinetobacter phage H	
2500136	Acinetobacter phage Henu6								EKI	>QAU04025.1 short tail fibers protein [Acinetobacter phage Henu6]M	
1735582	Acinetobacter phage IME20									9P0IJ73 Tail fiber protein [Acinetobacter phage IME200]MNILRS	
1243183	Acinetobacter pha									ail fiber protein-like protein [Acinetobacter phage IME	
1243183	Acinetobact									fiber [Acinetobacter phage IME-AB2]MTNPTLIT	
1243183	Aci									zyme protein [Acinetobacter phage I	
2301662										it [Acinetobacter phage KARL	
2301662										cinetobacter phage K	
				ATGGCGAC						ge KARL-1]M	
				AANGIKNL	GTGGCTAATCCTATT						MANPIFIP
				TLITTPFAENGDKN	ATGACAAATCCAACACT						5]MTNPTLIT
1				MAYVEKEMNLKRSFVEYK	ATGGCTTATGTAGAAAA						MAYVEKEMN
69			n fiber	MNILRSFTETVVTPTDTF	ATGAATATACTACGATC	882	Ac				B1]MNILRSFTE
169			n fiber	MNILRSFTETVVTPTDIFF	ATGAATATACTACGCTC	699	Acinetoba				]MNILRSFTETVV
1406			Lytic tail fiber	MARIDWQPLQAAPVNAS	ATGGCTCGTATTGATTGT	599	Acinetoba	MARIDW			y]MARIDWQPLQ
22021		47.1	tail fiber	MSKTLKIEVAIDDLDPNG	ATGTCAAAGACACTTAA	797	Acinetoba	MSKTLKIE	MSKTL		15599]MSKTLKIEV
1481186		A0A075DXC3	Putative tail fiber pro	MVAIVKPDFENIWASEGG	ATGGTAGCTATAGTAAA	359	Acinetoba	MVAIVKP	MVAIVKPDFE		bacter phage vB_Ab
1481186	aM_Acit	A0A075DXV3	Putative tail fiber pro	MSKVLTGNETVSIVGNFA	ATGTCAAAAGTATTAAC	402	Acinetoba	MSKVLTGN	MSKVLTGNETV	>A0A0	tobacter phage vB_Ab

Computationally Predicted and Validated RBPs after Processement with Python

3rd Dataset Analysis – K-means and PCA

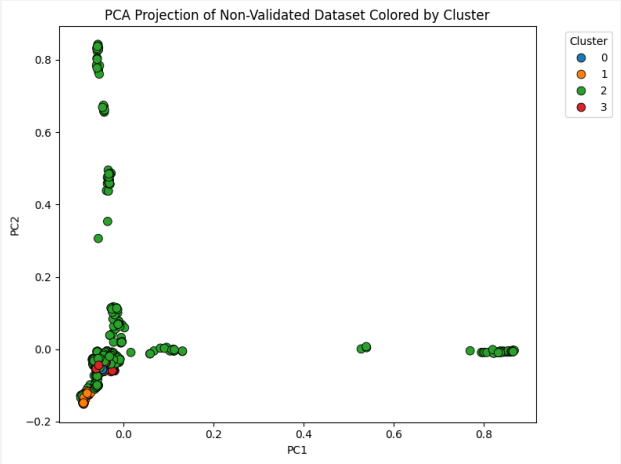
```python
import pandas as pd
from sklearn.cluster import KMeans
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.metrics import silhouette_score
from sklearn.decomposition import PCA
from sklearn.metrics import pairwise_distances
from scipy.cluster.hierarchy import dendrogram, linkage
import matplotlib.pyplot as plt
import seaborn as sns

# === Load dataset ===
df = pd.read_excel("phage_processed1.xlsx")

# === Vectorization ===
sequences = df["CleanSeq"].astype(str).tolist()
vectorizer = TfidfVectorizer(analyzer='char', ngram_range=(3, 4))
X = vectorizer.fit_transform(sequences)

# === Find optimal k using silhouette score ===
print("Finding best k based on silhouette score...")
best_k = None
best_score = -1
for k in range(2, 5):
    kmeans = KMeans(n_clusters=k, random_state=42, n_init='auto')
    labels = kmeans.fit_predict(X)
    score = silhouette_score(X, labels)
    print(f"k={k}, silhouette score={score:.4f}")
    if score > best_score:
        best_score = score
        best_k = k
print(f"Best k = {best_k} with silhouette score = {best_score:.4f}")

# === Final clustering with best k ===
kmeans = KMeans(n_clusters=best_k, random_state=42, n_init='auto')
df['Cluster'] = kmeans.fit_predict(X)
```

```
Finding best k based on silhouette score...
k=2, silhouette score=0.0140
k=3, silhouette score=0.0233
k=4, silhouette score=0.0301
Best k = 4 with silhouette score = 0.0301
```

```
Best params: {'n_clusters': 9, 'init': 'random'} with silhouette score: 0.0774
```

```python
# === PCA Visualization ===
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X.toarray())
df["PC1"] = X_pca[:, 0]
df["PC2"] = X_pca[:, 1]

plt.figure(figsize=(8, 6))
sns.scatterplot(data=df, x="PC1", y="PC2", hue="Cluster", palette="tab10", s=60, edgecolor='k')
plt.title("PCA Projection of Non-Validated Dataset Colored by Cluster")
plt.legend(title="Cluster", bbox_to_anchor=(1.05, 1), loc='upper left')
plt.tight_layout()
plt.show()

# === Dendrogram Visualization ===
distance_matrix = pairwise_distances(X.toarray(), metric='cosine')

Z = linkage(distance_matrix, method='ward')

plt.figure(figsize=(30, 10))
labels = df.apply(lambda row: f"Cl{row['Cluster']}_{row['RBP_Accession']}", axis=1).values
dendrogram(Z, labels=labels, truncate_mode='level', p=10, leaf_rotation=90)
plt.xticks(rotation=90, size=8)
plt.title('Dendrogram of Both RBPs Clustering')
plt.xlabel('Sample Index or Cluster ID')
plt.ylabel('Distance')
plt.show()


# === Save results ===
df.to_excel("rbp_clustered_results_unsupervised.xlsx", index=False)
print("\nSaved results to 'rbp_clustered_results_unsupervised.xlsx'")
```

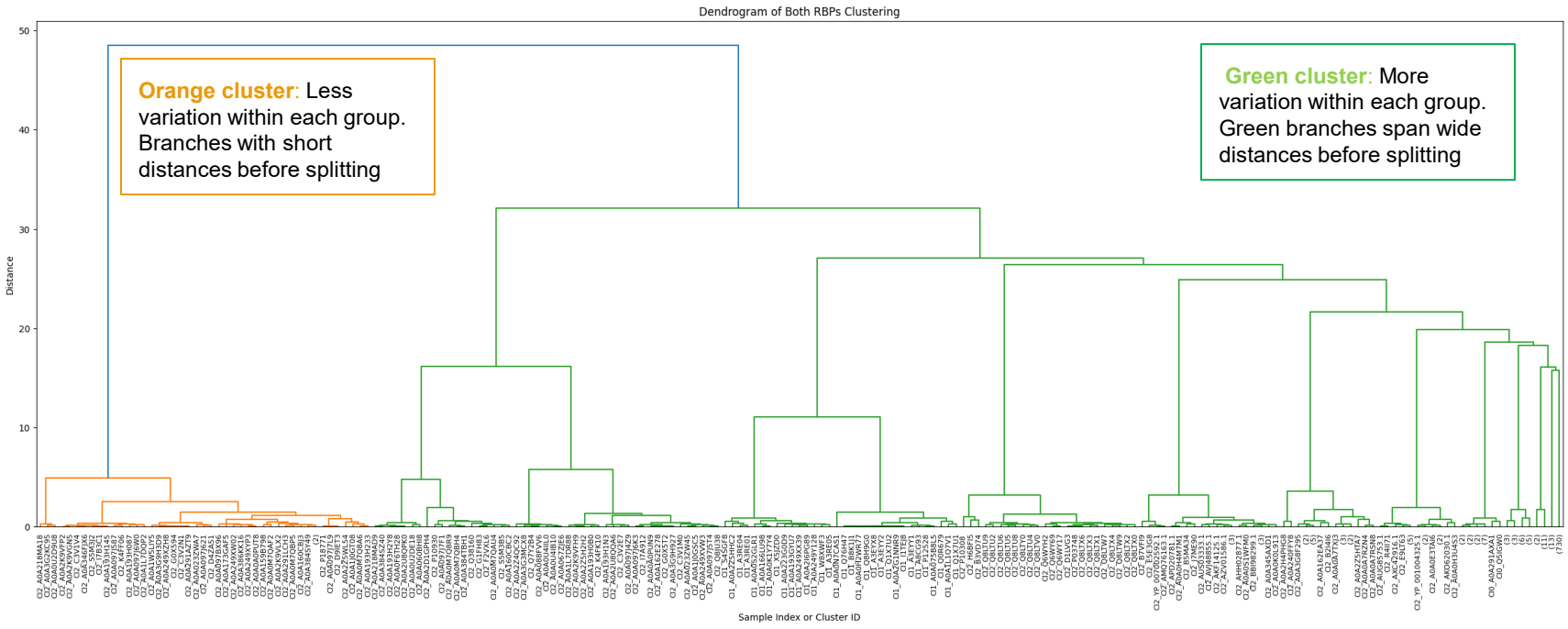
```python
best_score = -1
best_params = {}

for k in range(2, 10):
    for init_method in ['k-means++', 'random']:
        model = KMeans(n_clusters=k, init=init_method, random_state=42, n_init='auto')
        labels = model.fit_predict(X)
        score = silhouette_score(X, labels)
        if score > best_score:
            best_score = score
            best_params = {'n_clusters': k, 'init': init_method}

print(f"Best params: {best_params} with silhouette score: {best_score:.4f}")
```



Cluster 0:
Numerous points but overlaps with **Cluster 2**
Cluster 2:
Has most points → dominant cluster in this dataset
Cluster 1, Cluster 3:
Less populated and overlap → similarity in characteristics

Interpretation:
- Distinct clusters → Validated and Non-Validated RBPs may have different characteristics

3rd Dataset Analysis – Dendrogram



- Cluster shape: Broad, with higher internal variation
- Separation between clusters: High
- Cohesion between sub-groups
- Nº Clusters: Low

- Dissimilar sequences that share functional elements

These clusters likely represent:

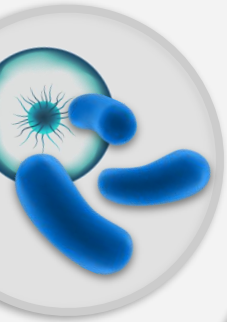
- Possibly novel or divergent RBPs
- Groupings of functionally analogous

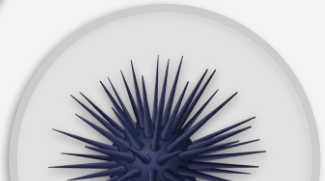
3rd Dataset – Both RBPs CD-HIT

```python
from Bio import SeqIO
import pandas as pd
def parse_cdhit_output(file_path):
    clusters = {}
    current_cluster = None

    with open(file_path, 'r') as file:
        for line in file:
            line = line.strip()
            if line.startswith(">Cluster"):
                current_cluster = line.split(" ")[1]
                clusters[current_cluster] = []
            elif current_cluster is not None and line:
                if '>' in line:
                    try:
                        seq_info = line.split(",")[1] if ',' in line else line
                        seq_id_part = seq_info.split(">")[1]
                        seq_id = seq_id_part.split("...")[0].strip()
                        if seq_id:
                            clusters[current_cluster].append(seq_id)
                    except Exception as e:
                        print(f"Error parsing line: {line}\n{e}")
    return clusters

def load_fasta_sequences(fasta_path):
    seq_dict = {}
    for record in SeqIO.parse(fasta_path, "fasta"):
        core_id = record.id
        seq_dict[core_id] = str(record.seq)
    return seq_dict
```

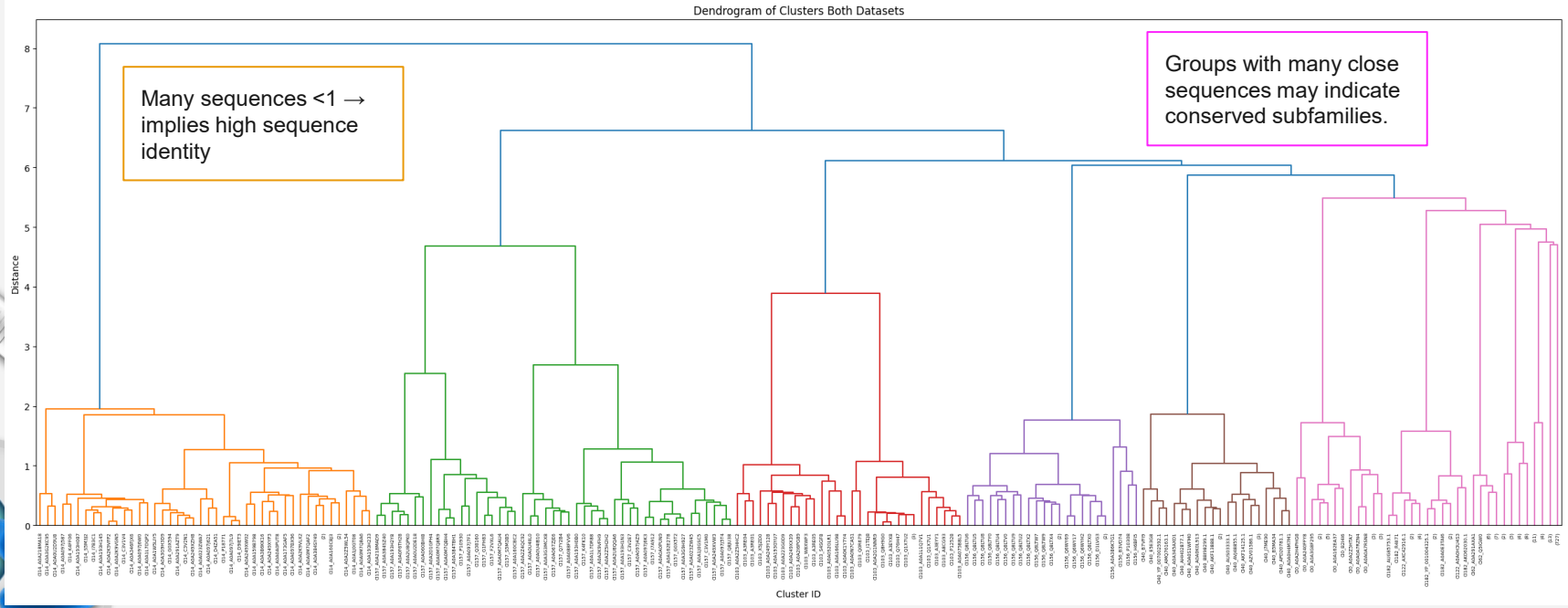
```python
def build_sequences_list(clusters, seq_dict):
    sequences = []
    data = []
    for cluster_id, seq_ids in clusters.items():
        for seq_id in seq_ids:
            seq = seq_dict.get(seq_id, "")
            if seq == "":
                print(f"Warning: Sequence ID '{seq_id}' not found in FASTA file.")
            data.append({'Cluster': cluster_id, 'Sequence ID': seq_id})
            sequences.append(seq)
    df = pd.DataFrame(data)
    return df, sequences

if __name__ == "__main__":
    cluster_file = "cluster_both.txt"
    fasta_file = "unique_phage_proteins1.fasta"

    clusters = parse_cdhit_output(cluster_file)
    seq_dict = load_fasta_sequences(fasta_file)
    df, sequences = build_sequences_list(clusters, seq_dict)

    print(f"Total sequences parsed: {len(sequences)}")
    df.to_csv("clustered_sequences1.csv", index=False)
```

```python
from scipy.cluster.hierarchy import dendrogram, linkage
import numpy as np
from sklearn.feature_extraction.text import TfidfVectorizer

vectorizer = TfidfVectorizer(analyzer='char', ngram_range=(3, 4))
X = vectorizer.fit_transform(sequences)
Z = linkage(X.toarray(), method='ward')

plt.figure(figsize=(30, 10))
labels = df.apply(lambda row: f"Cl{row['Cluster']}_{row['Sequence ID']}", axis=1).values
dendrogram(Z, labels=labels, truncate_mode='level', p=10, leaf_rotation=90)
plt.xticks(rotation=90, size=5)
plt.title('Dendrogram of Clusters Both Datasets')
plt.xlabel('Cluster ID')
plt.ylabel('Distance')
plt.show()
```





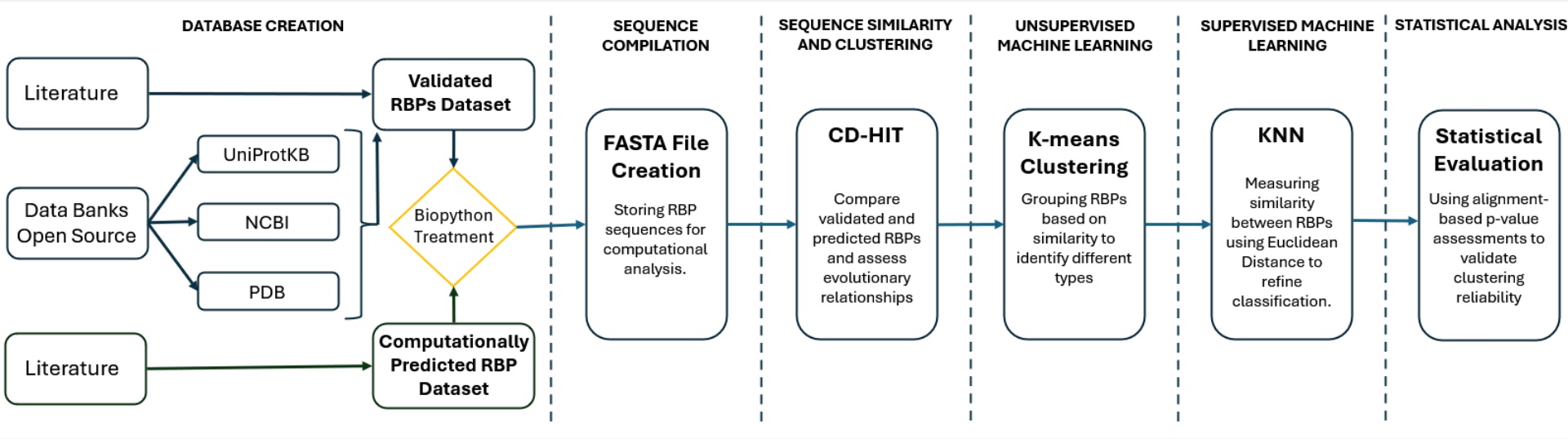
3rd Dataset – Both RBPs CD-HIT



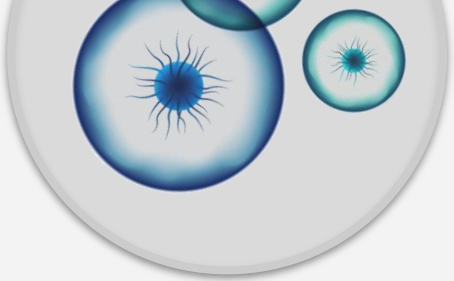
- Cluster shape: Compact, narrow (short branches within)
- Separation between clusters: High (clear distinction)
- Intra-cluster similarity: High (identity ≥ 40%)
- Inter-cluster difference: High (hard cutoff)

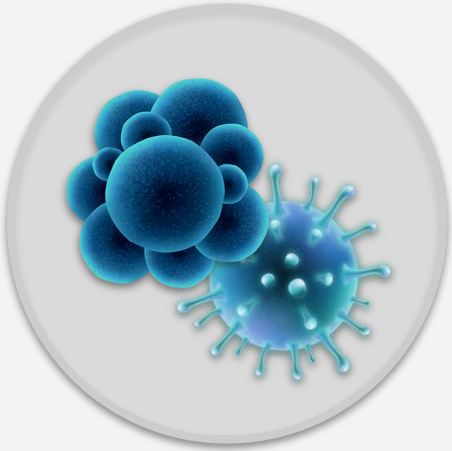
Each cluster could correspond to:

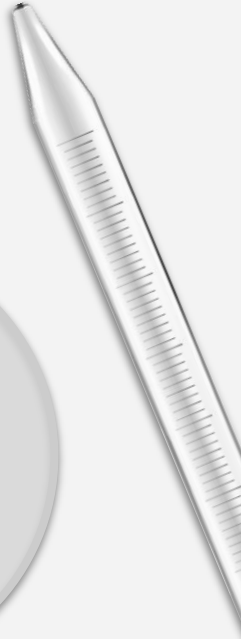
- One phage RBP subtype,
- A known structural domain or tail fiber class,
- Evolutionary lineages with strong sequence conservation.



PROJECT WORKFLOW







05

Discussion

Discussion

Challenges

Creation of the provisory datasets:
- Filter and selection of RBPs from the 3 Databases
- Automatization of morphotype retrieval
- Datasets Uniformization

Validation of RBPs:
- Literature key word identification
- Explicit experimental data

Success

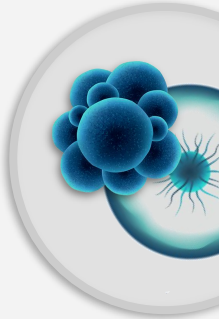
Curated RBP Database:
- Two datasets with Validated RBP and RBP candidates

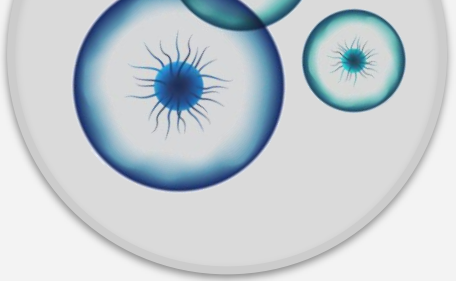
Implemented dual clustering approaches:

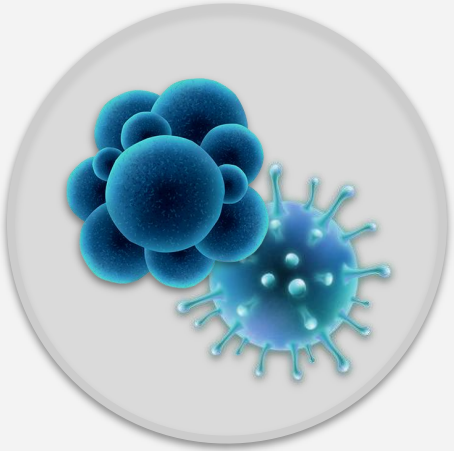
- CD-HIT (identity-based): Captured conserved structural/evolutionary groups

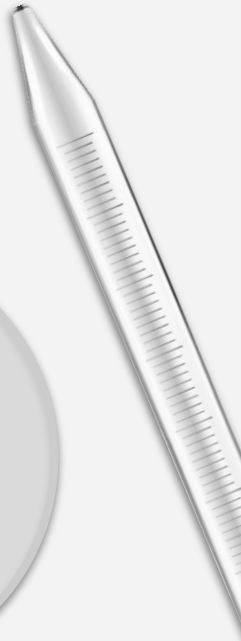
- K-Means (motif-based): Revealed functional similarity based on k-mer features

Visualized clustering via dendrograms:
- Assess sequence relationships









06

Conclusion

Conclusion

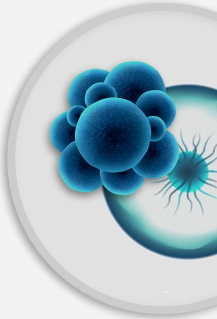
Next Steps

- **Propythia -** platform for the classification of biological sequences (proteins and DNA) using machine and deep learning

- Train classifiers using validated clusters as labels
- Deploy the RBP-ID web Based tool
- Enable user input and visualization, returning predicted RBP status, cluster info, and motifs
- Expand the validated dataset via literature mining

Vision

RBP-ID aims to become a go-to platform for phage researchers, accelerating discovery of receptor-binding proteins for synthetic biology, diagnostics, and phage therapy



Thanks