

WHAT MAKES EXTREMOPHILES SPECIAL?

Comprehensive Analysis of *Halobacterium* sp. *GSL-19*

BMI 6015 Foundations in Bioinformatics, University of Utah - Fall 2024
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Objective

Our exploration aims to gain insights into the molecular mechanisms of survival in harsh conditions.

To do this, we perform a comprehensive analysis of *Halobacterium sp. GSL-19*, an organism that exhibits adaptations to extreme salinity and alkalinity.

Domain: Archaea

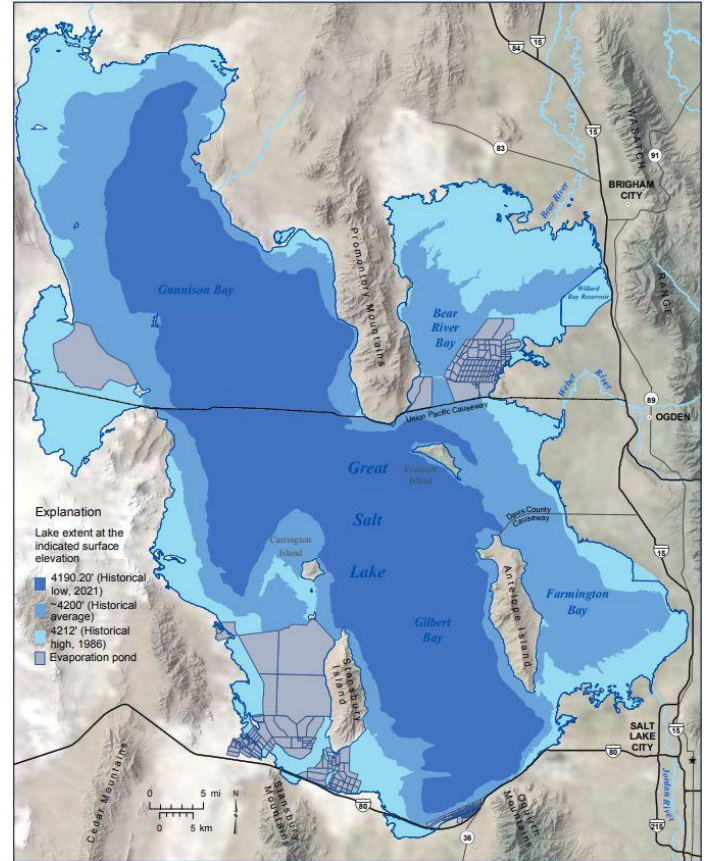
Class: Halobacteria

Genus: *Halobacterium*

Species: *Halobacterium sp. GSL-19*

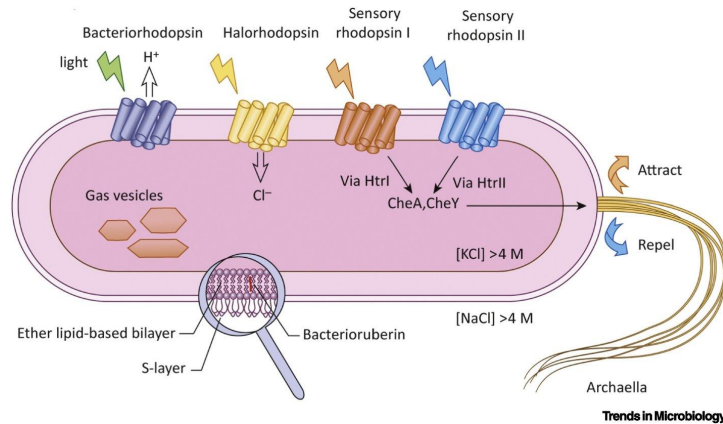
Origin: **North arm of Great Salt Lake**, Utah (41.4377°N, 112.6689°W)

Environmental Conditions: **Hypersaline** (≥ 300 g/L NaCl), alkaline pH



Introduction

Halobacteria are halophilic microorganisms, which means they grow in extremely high salinity environments. Comparing a halophile genome to that of other prokaryotes should give insight into microbial adaptation to extreme conditions.



Background: “what makes halophiles special?”

Compact genomes and specialized salt tolerance mechanisms.



- "Salt-in" strategy, where intracellular potassium compensates for external salinity, reducing osmotic stress
- Specialized protein folding pathways and acidic proteomes for structural integrity
- Modified membrane lipids enhance stability under hypersaline and alkaline conditions
- Adapted to high UV environments

2 Analyzes using GSL-19

1. **Halobacterium sp. GSL-19:** extreme halophile located in the north arm of the *Great Salt Lake*.
2. **Halanaerobium praevalen:** moderate halophile located the north arm of the *Great Salt Lake*.
3. **Halobacterium sp. NRC-1:** common halophile reference genome.
4. **Halobacterium salinarum 91 R-6:** GenBank official halophile reference genome.
5. **Halobacterium sp. NMX12-1:** ancient halophile from the Permian Period found in a salt formation in *New Mexico*.
6. **Escherichia coli:** non-halophile.

Analysis 1 with Jupyter notebooks and Galaxy

Goal: comprehensive content and feature analysis

Step 1: Sequence analysis

```
def download_genome_sequence(accession):
    """Download genome sequence from GenBank"""
    handle = Entrez.efetch(db="nucleotide", id=accession, rettype="gb", retmode="text")
    return SeqIO.read(handle, "genbank")

# Download the three genome sequences
accessions = ["CP070375.1", "CP070376.1", "CP070377.1", "U00096", "CP002175.1"]
genome_records = {}

for acc in accessions:
    genome_records[acc] = download_genome_sequence(acc)
    print(f"Downloaded sequence: {acc}")

# Analyze each sequence
for acc, record in genome_records.items():
    print(f"\nAnalyzing sequence: {acc}")
    print(f"Sequence length: {len(record.seq)} bp")
    print(f"GC content: {(record.seq.count('G') + record.seq.count('C')) / len(record.seq) * 100:.2f}%")
    print(f"Number of features: {len(record.features)}")
```


GSL-19 Plasmid 1 →

Analyzing sequence: CP070375.1
Sequence length: 284178 bp
GC content: 59.07%
Number of features: 549

GSL-19 Plasmid 2 →

Analyzing sequence: CP070376.1
Sequence length: 54914 bp
GC content: 61.37%
Number of features: 122

GSL-19 Chromosome →

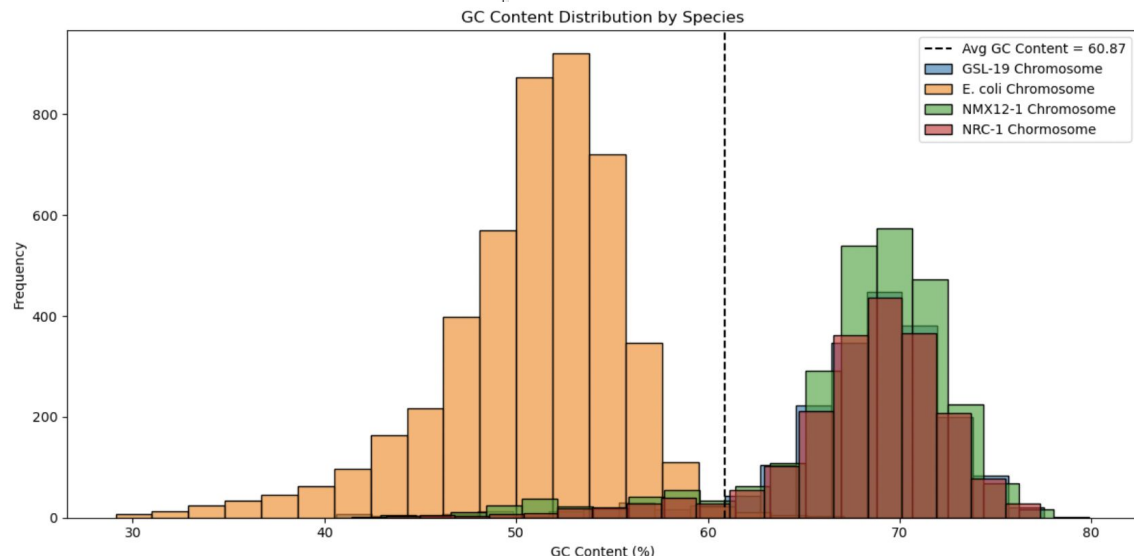
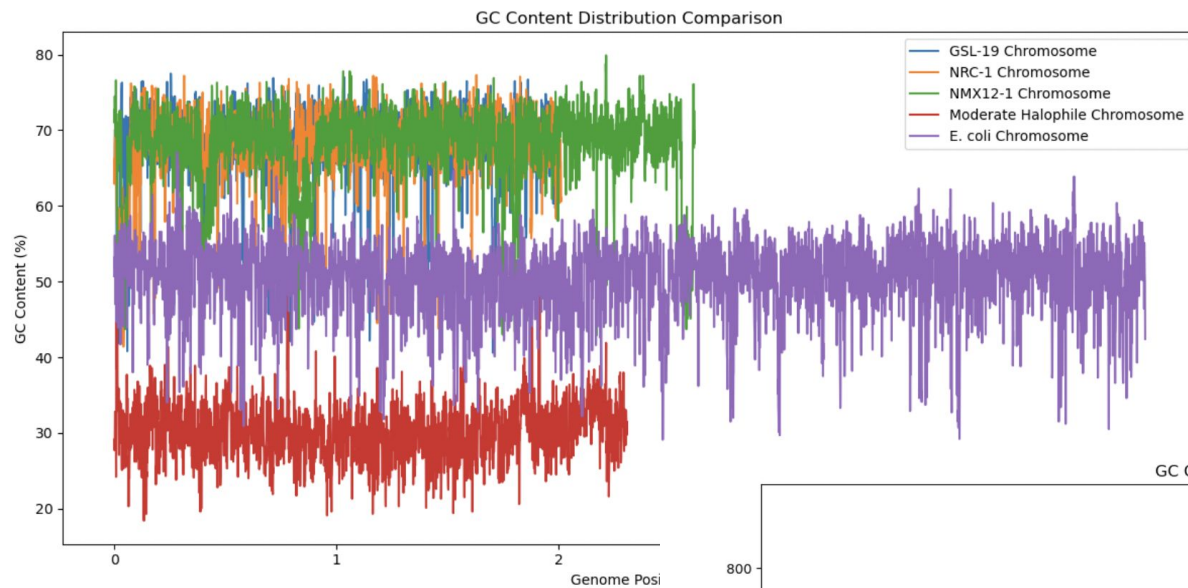
Analyzing sequence: CP070377.1
Sequence length: 1987132 bp
GC content: 67.99%
Number of features: 4247

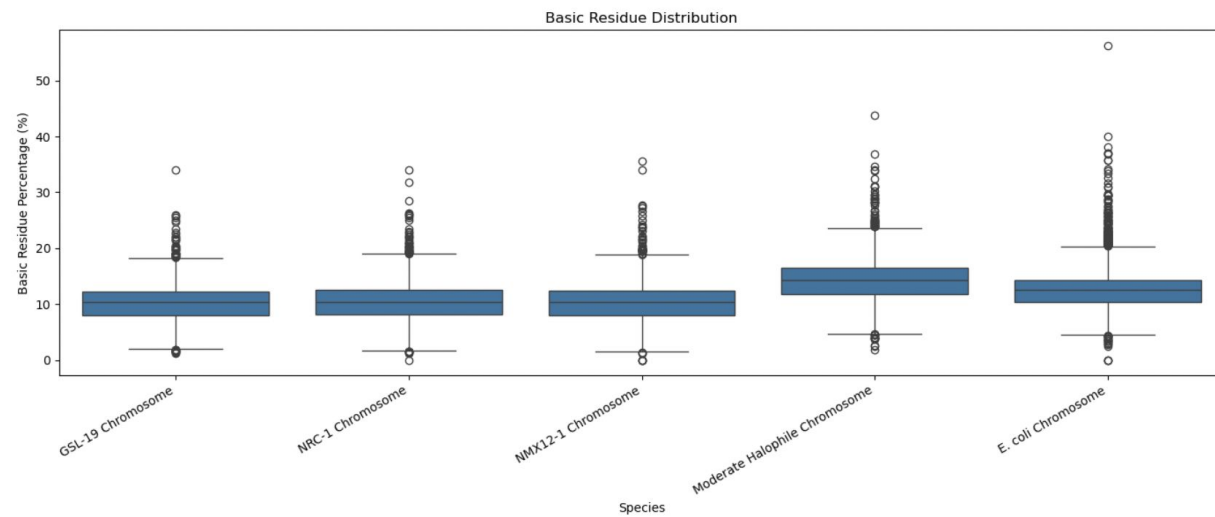
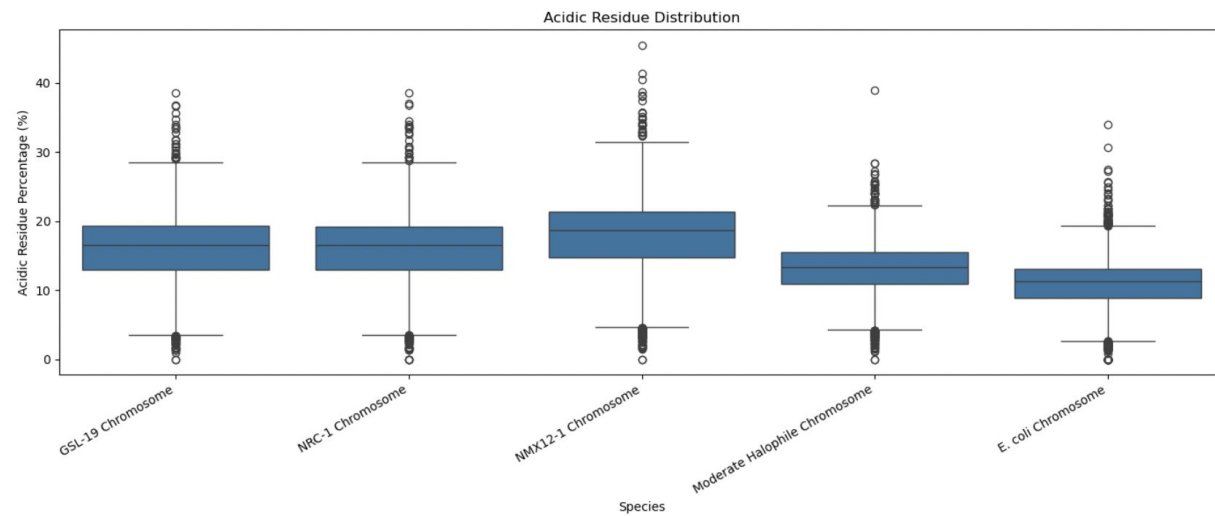
E coli. →

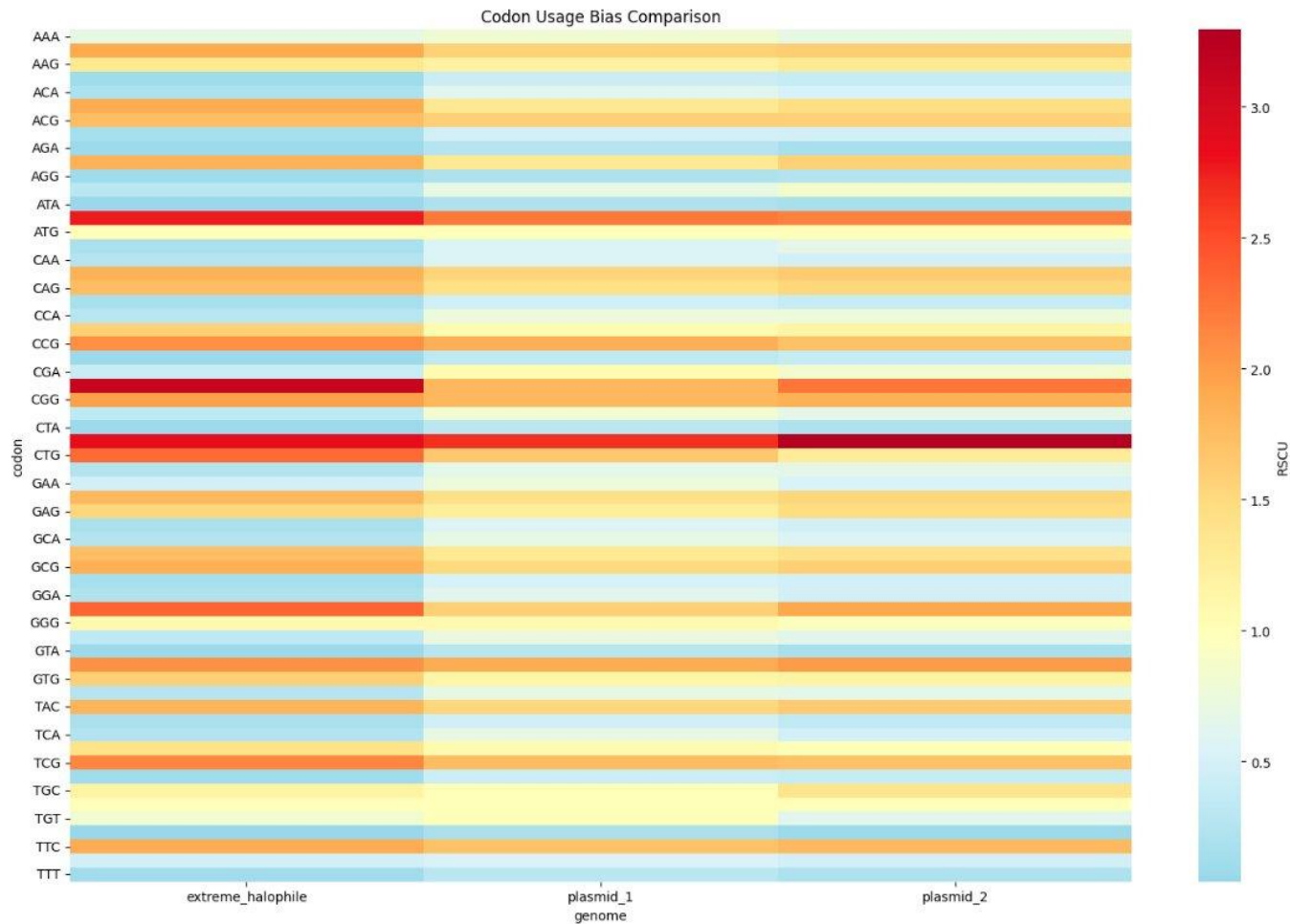
Analyzing sequence: U00096
Sequence length: 4641652 bp
GC content: 50.79%
Number of features: 9285

H praevalens →

Analyzing sequence: CP002175.1
Sequence length: 2309262 bp
GC content: 30.29%
Number of features: 4875







Step 2: Feature analysis

```
def analyze_features(record):  
    """Analyze features in the genome record"""  
    feature_types = {}  
    for feature in record.features:  
        if feature.type in feature_types:  
            feature_types[feature.type] += 1  
        else:  
            feature_types[feature.type] = 1  
    return feature_types  
  
# Create feature analysis for each sequence  
for acc, record in genome_records.items():  
    print(f"\nFeature analysis for {acc}:")  
    features = analyze_features(record)  
    for feature_type, count in features.items():  
        print(f"{feature_type}: {count}")
```

163: Prokka on data 127: ffn

162: Prokka on data 127: faa

161: Prokka on data 127: fna

160: Prokka on data 127: gbk

159: Prokka on NMX12-1: gff

158: Prokka on data 126: log

157: Prokka on data 126: txt

156: Prokka on data 126: tsv

155: Prokka on data 126: tbl

154: Prokka on data 126: gff

153: Prokka on data 126: faa

152: Prokka on data 126: fna

151: Prokka on data 126: gbk

150: Prokka on data 126: log

149: Prokka on data 126: txt

194: Interactive JupyterLab Notebook on data 191

193: Executed Jupyter Notebook

192: Jupyter output collection on

a list with 0 datasets

```
organism: Genus species strain  
contigs: 3  
bases: 2326224  
CDS: 2401  
gene: 2452  
rRNA: 3  
repeat_region: 4  
tRNA: 48
```

GSL-19 Plasmid 1 →

GSL-19 Plasmid 2 →

GSL-19 Chromosome →

E coli. →

H praevale n →

```
Feature analysis for CP070375.1:  
source: 1  
gene: 274  
CDS: 274
```

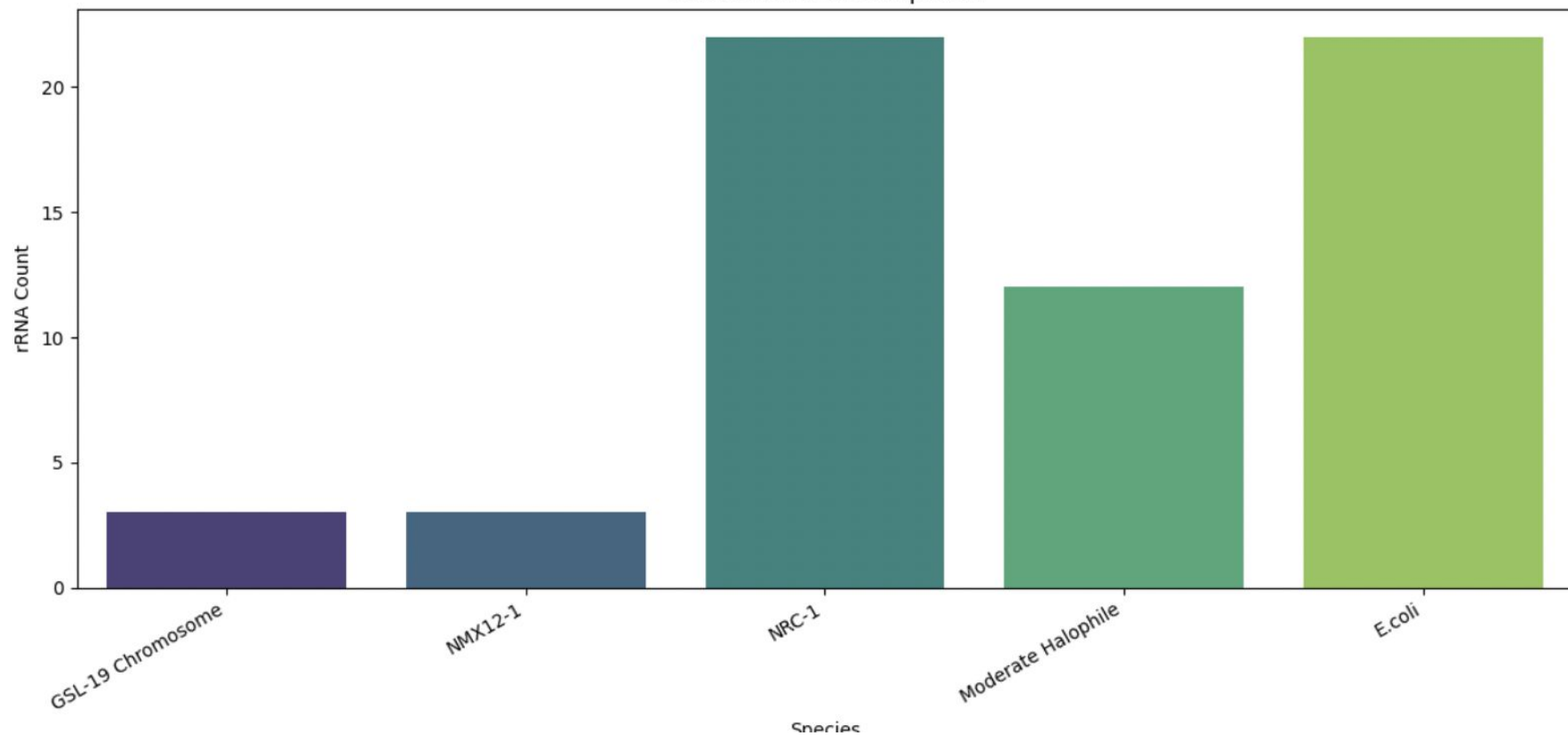
```
Feature analysis for CP070376.1:  
source: 1  
gene: 59  
CDS: 59  
repeat_region: 3
```

```
Feature analysis for CP070377.1:  
source: 1  
gene: 2123  
CDS: 2071  
tRNA: 47  
ncRNA: 2  
rRNA: 3
```

```
Feature analysis for U00096:  
source: 1  
gene: 4651  
CDS: 4318  
mobile_element: 50  
...  
tRNA: 55  
sig_peptide: 555  
ncRNA: 3  
repeat_region: 1
```

```
Feature analysis for CP002175.1:  
source: 1  
gene: 2180  
CDS: 2068  
rRNA: 12  
tRNA: 55  
sig_peptide: 555  
ncRNA: 3
```


rRNA Counts Across Species



Step 3: Gene extraction

```
def extract_relevant_genes(genome_record):
    """Extract genes related to stress response, transporters, or gas vesicle proteins."""
    relevant_genes = []
    # Keywords to identify relevant genes
    keywords = ["stress", "gas vesicle", "potassium", "transport", "haloarchaea", "hypersaline", "ABC transporter"]

    for feature in genome_record.features:
        if feature.type == "CDS": # Look at coding sequences
            product = feature.qualifiers.get("product", [""])[0].lower()
            if any(keyword in product for keyword in keywords):
                locus_tag = feature.qualifiers.get("locus_tag", ["N/A"])[0]
                relevant_genes.append({
                    "locus_tag": locus_tag,
                    "product": product
                })
    return relevant_genes

# Extract genes matching the updated criteria
all_relevant_genes = {}

for acc, record in genome_records.items():
    relevant_genes = extract_relevant_genes(record)
    all_relevant_genes[acc] = relevant_genes
    print(f"Identified {len(relevant_genes)} relevant genes in {acc}.")

# Print the identified genes
for acc, genes in all_relevant_genes.items():
    print(f"\nAccession: {acc}")
    for gene in genes:
        print(f"Locus Tag: {gene['locus_tag']}, Product: {gene['product']}")
```

We found many instances of osmotic balance genes, protein adaptations, and membrane adaptations in GSL-19.

Identified 14 gas vesicle-related genes in CP070375.1.

Identified 0 gas vesicle-related genes in CP070376.1.

Identified 0 gas vesicle-related genes in CP070377.1.

Accession: CP070375.1

Locus Tag: JT689_00385, Product: gas vesicle protein

Locus Tag: JT689_00390, Product: gas vesicle protein gvpl

Locus Tag: JT689_00395, Product: gas vesicle protein k

Locus Tag: JT689_00400, Product: gas vesicle protein gvpj

Locus Tag: JT689_00405, Product: gas vesicle protein gvpi

Locus Tag: JT689_00410, Product: gas vesicle protein gvph

Locus Tag: JT689_00415, Product: gas vesicle protein gvpg

Locus Tag: JT689_00420, Product: gas vesicle protein gvpf

Locus Tag: JT689_00425, Product: gas vesicle transcriptional activator gvpe

Locus Tag: JT689_00430, Product: gas vesicle protein gvpd

Locus Tag: JT689_00435, Product: gas vesicle structural protein gvpa

Locus Tag: JT689_00440, Product: gas vesicle protein gvpc

Locus Tag: JT689_00445, Product: gas vesicle protein gvpn

Locus Tag: JT689_00450, Product: gas vesicle protein gvpo

Gas vesicle proteins

Gas vesicles are a **crucial adaptation** in halophilic archaea that allow these microorganisms to regulate their buoyancy in hypersaline environments.

Gvp's can be used as contrast agents, delivery carriers, and immunology boosters for disease prevention, diagnosis, and treatment. Largely due to their tiny size, strong stability and non-toxic advantages.

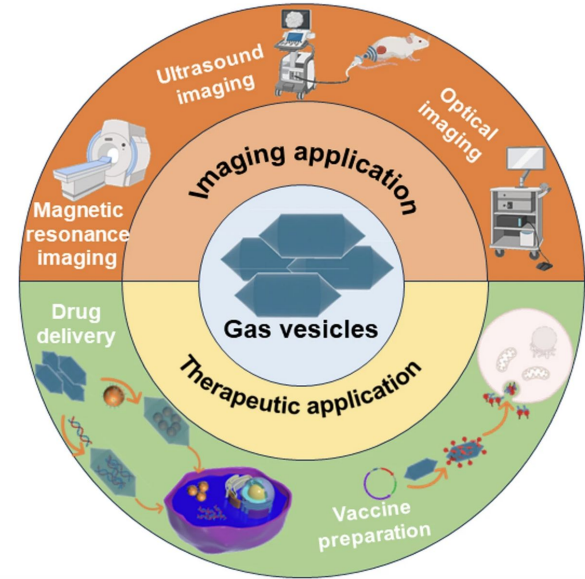


Figure 1 (Feng et al. 2024), (Jost and Pfeifer 2022)

Step 4: Analysis of gas vesicle proteins using BLAST

Pipeline Overview:

1. Define target gas vesicle loci and analysis parameters
2. Extract genomic regions surrounding gas vesicle genes
3. Perform pairwise sequence alignments
4. Conduct BLAST analysis for functional annotation
5. Analyze synteny and conservation patterns

```
# Step 1: Define gas vesicle locus tags and flanking region size
gas_vesicle_loci = ["JT689_00385", "JT689_00390", "JT689_00395", "JT689_00400",
                   "JT689_00405", "JT689_00410", "JT689_00415", "JT689_00420",
                   "JT689_00425", "JT689_00430", "JT689_00435", "JT689_00440",
                   "JT689_00445", "JT689_00450"]

flank_size = 10000 # 10 kb

# Step 2: Extract genomic regions surrounding gas vesicle genes
def extract_flanking_regions(genome_record, loci, flank_size):
    regions = {}
    for feature in genome_record.features:
        if feature.type == "CDS" and feature.qualifiers.get("locus_tag", [None])[0] in loci:
            start = max(0, feature.location.start - flank_size)
            end = min(len(genome_record.seq), feature.location.end + flank_size)
            regions[feature.qualifiers["locus_tag"][0]] = genome_record.seq[start:end]
    return regions

# Fetch genome record for Halobacterium sp. GSL-19 (example for one chromosome)
halobacterium_record = genome_records["CP070375.1"]
flanking_regions = extract_flanking_regions(halobacterium_record, gas_vesicle_loci, flank_size)

# Step 3: Pairwise alignment example
for locus, region in flanking_regions.items():
    print(f"Analyzing locus: {locus}")
    # Example: Compare to E. coli genome
    ecoli_region = genome_records["U00096"].seq[:len(region)] # Truncated for simplicity
    alignments = pairwise2.align.globalxx(region, ecoli_region)
    print(f"Alignment score for {locus}: {alignments[0].score}")

# Step 4: Functional annotation (use BLAST for alignment to databases)
for locus, region in flanking_regions.items():
    print(f"Running BLAST for locus: {locus}")
    result_handle = NCBIWWW.qblast("blastn", "nt", str(region))
    blast_record = NCBIXML.read(result_handle)
    for alignment in blast_record.alignments[:5]: # Top 5 matches
        print(f"Match: {alignment.title}, Score: {alignment.hsps[0].score}")
```

Match: gi|2707227173|gb|CP128378.3| Halobacterium salinarum strain KBTZ01 plasmid pKBTZ01_286, complete sequence, Score: 40672.0
Match: gi|1992388726|gb|CP070375.1| Halobacterium sp. GSL-19 plasmid pGSL19_284, complete sequence, Score: 40672.0
Match: gi|2696630081|gb|CP146626.1| Halobacterium salinarum strain KBTZ03 plasmid pKBTZ03_355, complete sequence, Score: 39984.0
Match: gi|164521090|gb|EU080936.1| Halobacterium sp. GN101 plasmid megaplasmid 2, complete sequence, Score: 39790.0

```
Analyzing locus: JT689_00385
Alignment score for JT689_00385: 13056.0
Analyzing locus: JT689_00390
Alignment score for JT689_00390: 13454.0
Analyzing locus: JT689_00395
Alignment score for JT689_00395: 13156.0
Analyzing locus: JT689_00400
Alignment score for JT689_00400: 13122.0
Analyzing locus: JT689_00405
Alignment score for JT689_00405: 13155.0
Analyzing locus: JT689_00410
Alignment score for JT689_00410: 13245.0
Analyzing locus: JT689_00415
Alignment score for JT689_00415: 13099.0
Analyzing locus: JT689_00420
Alignment score for JT689_00420: 13370.0
Analyzing locus: JT689_00425
Alignment score for JT689_00425: 13308.0
Analyzing locus: JT689_00430
Alignment score for JT689_00430: 13873.0
Analyzing locus: JT689_00435
Alignment score for JT689_00435: 13063.0
Analyzing locus: JT689_00440
Alignment score for JT689_00440: 13475.0
Analyzing locus: JT689_00445
...
```


BLAST hits reveal interesting patterns: most matches are to plasmid sequences, with a consistent presence on large plasmids. Suggests potential for horizontal gene transfer.

1. GSL-19 (consistently highest scores)
2. *H. salinarum* strains KBTZ02 and KBTZ01 (very similar scores)
3. *H. salinarum* strain KBTZ03 (slightly lower scores)
4. *Halobacterium* sp. GN101 (consistently lowest scores)

Analysis 2 Genome Comparison in Galaxy

Goal: compare gas vesicle protein genes across
6 bacteria species/strains

Gas vesicle protein variations

GvpA and C are particularly important as they (respectively) form the ribs of the vesicle and stabilize the structure, while **gvpN and O** are found in lower amounts.

The functions of **gvpEFGHIJKLM** are not completely understood, but some are required.

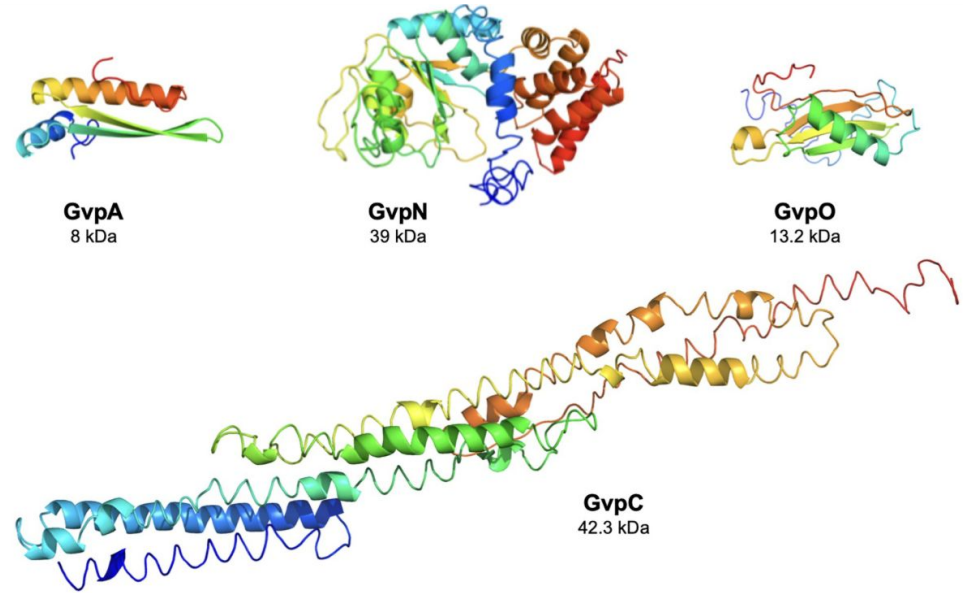
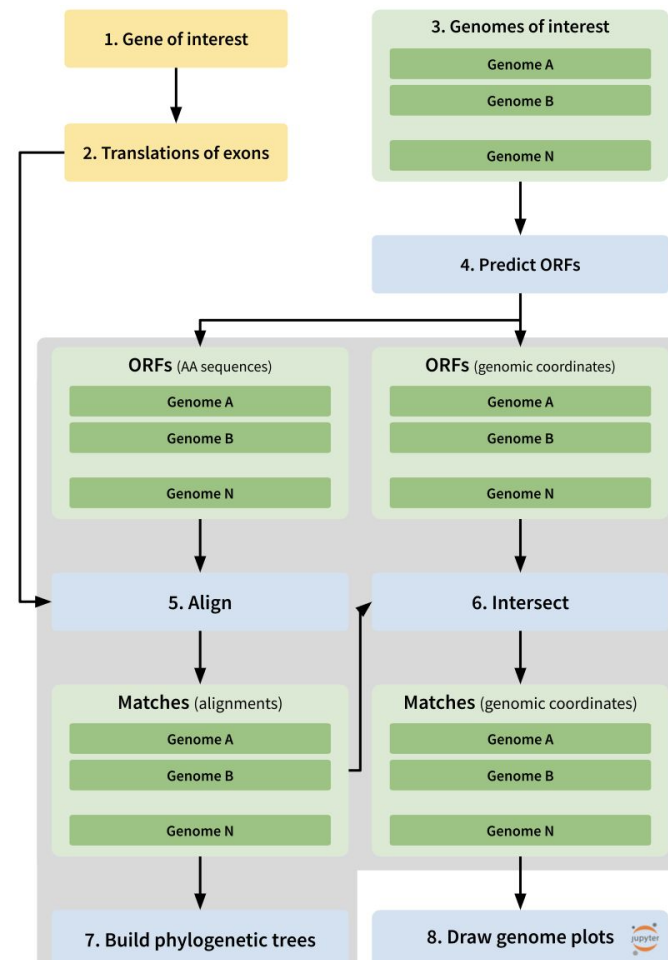


Figure 1 (Jost and Pfeifer 2022)

Comparative Gene Analysis Workflow

1. upload **gvpA** amino acid translation
2. upload **the 6 species** genomes as a dataset
3. extract ORF's from the genomes
4. create a diamond database
5. run tutorial workflow
6. Jupyter notebooks visuals

Output: phylogenetic trees and gene location graphs



Step 1 and 2: Upload data

14: gvpA_protein.faa

13: dataset collection

a list with 6 **fasta** datasets

dataset collection

a list with 6 **fasta** datasets

1: 91R6_GCF_004799605.1_ASM479960v1.fna

2: NRC1_GCF_000006805.1_ASM680v1.fna

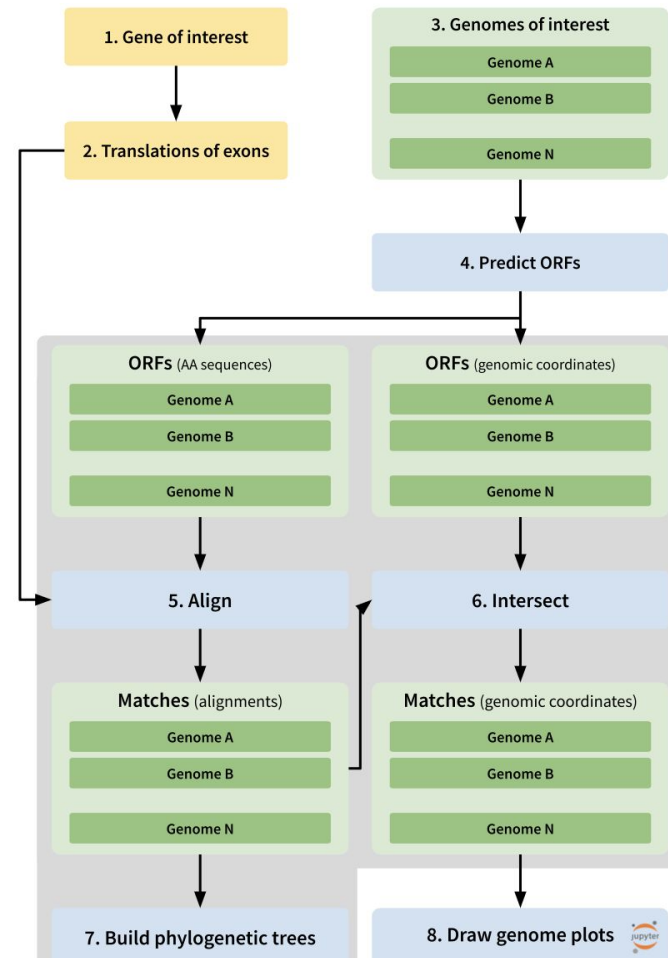
3: NMX12_GCF_040435975.1_ASM4043597v1.fna

4: Hprev_GCF_000165465.1_ASM16546v1.fna

5: GSL19_GCF_016917855.1_ASM1691785v1.fna

6: Ecoli_GCF_000005845.2_ASM584v2.fna

```
>WP_010904108.1 MULTISPECIES: gas vesicle protein GvpA [Halobacterium]  
MAQPDSSSLAEVLDRVLDKGVVDVVARISLVGIEILTVEARVVAASVDTLHYAEEIAKIEQAELTAGA  
EAEPEAPEA
```



Step 3: ORFiPy

ORFs on collection 13 (BED format)

a list with 6 **bed6** datasets

1: 91R6_GCF_004799605.1_ASM479960v1.fna

2: NRC1_GCF_000006805.1_ASM680v1.fna

3: NMX12_GCF_040435975.1_ASM4043597v1.fna

4: Hprev_GCF_000165465.1_ASM16546v1.fna

5: GSL19_GCF_016917855.1_ASM1691785v1.fna

6: Ecoli_GCF_000005845.2_ASM584v2.fna

ORFs on collection 13 (FASTA Protein)

a list with 6 **fasta** datasets

1: 91R6_GCF_004799605.1_ASM479960v1.fna

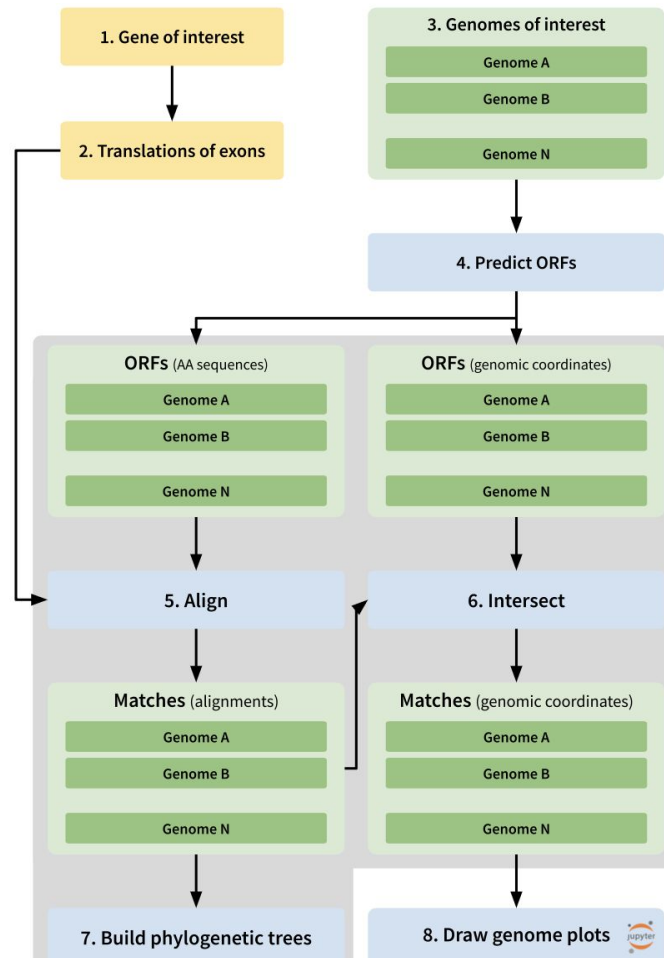
2: NRC1_GCF_000006805.1_ASM680v1.fna

3: NMX12_GCF_040435975.1_ASM4043597v1.fna

4: Hprev_GCF_000165465.1_ASM16546v1.fna

5: GSL19_GCF_016917855.1_ASM1691785v1.fna

6: Ecoli_GCF_000005845.2_ASM584v2.fna



Step 4: Diamond makedb

Diamond makedb on collection 16

a list with 6 **dmnd** datasets

1: 91R6_GCF_004799605.1_ASM479960v1.fna



2: NRC1_GCF_000006805.1_ASM680v1.fna



3: NMX12_GCF_040435975.1_ASM4043597v1.fna



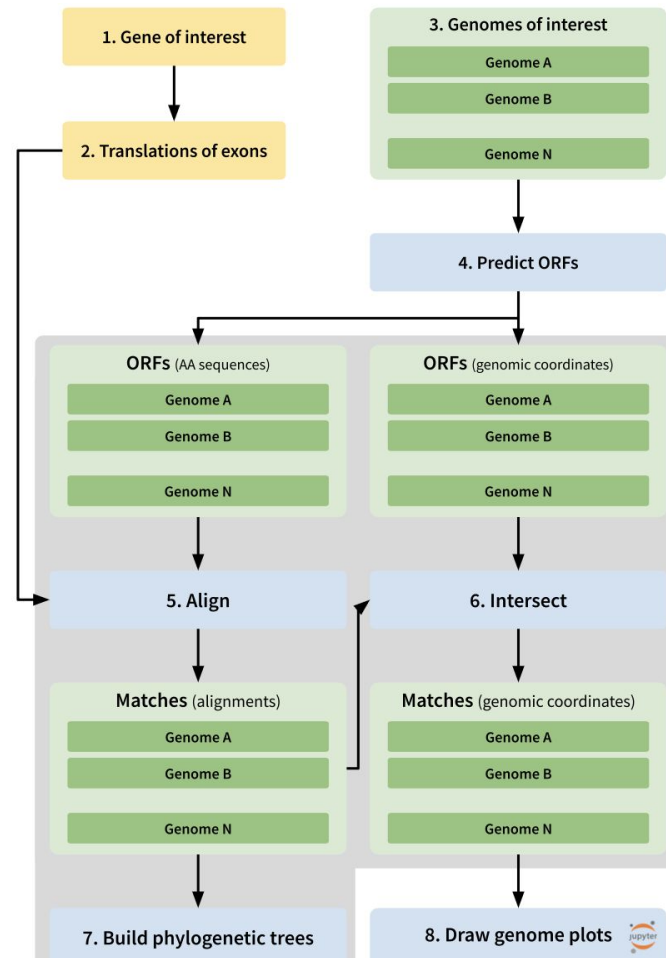
4: Hprev_GCF_000165465.1_ASM16546v1.fna



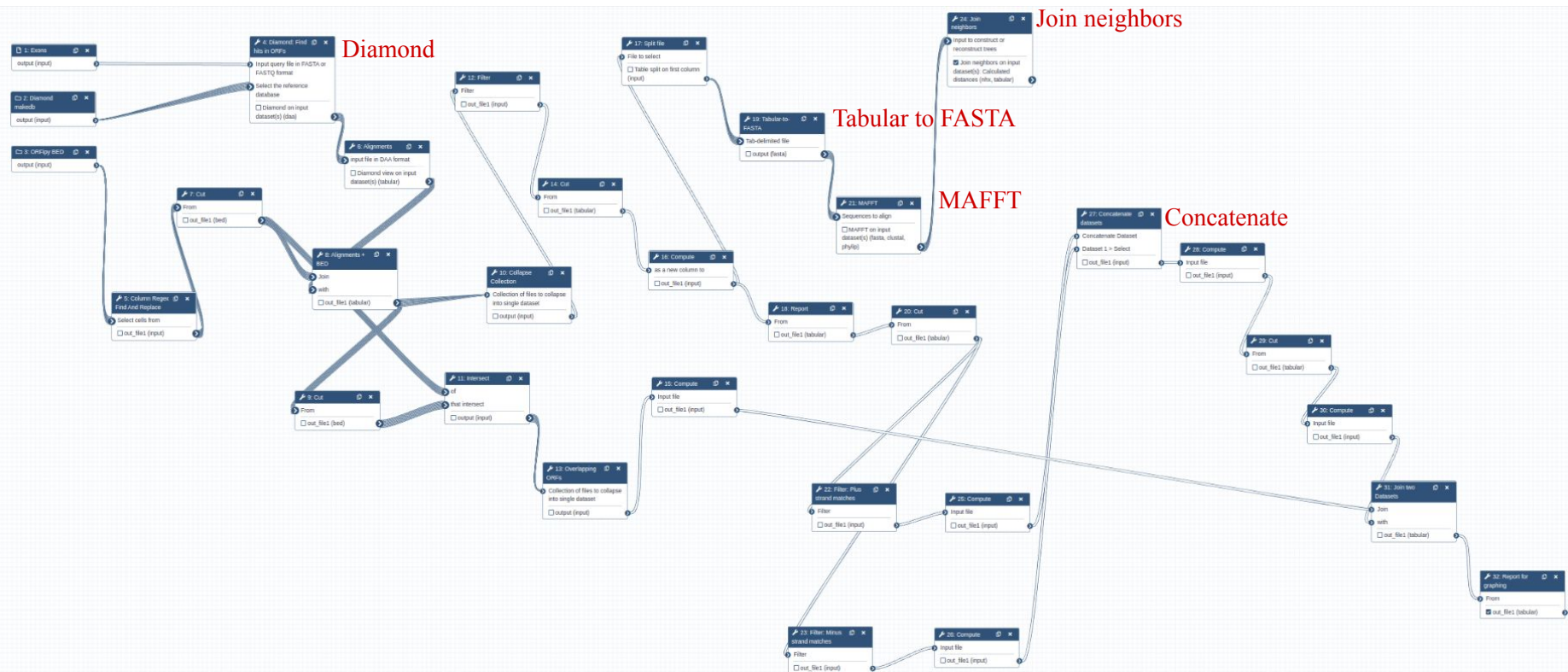
5: GSL19_GCF_016917855.1_ASM1691785v1.fna

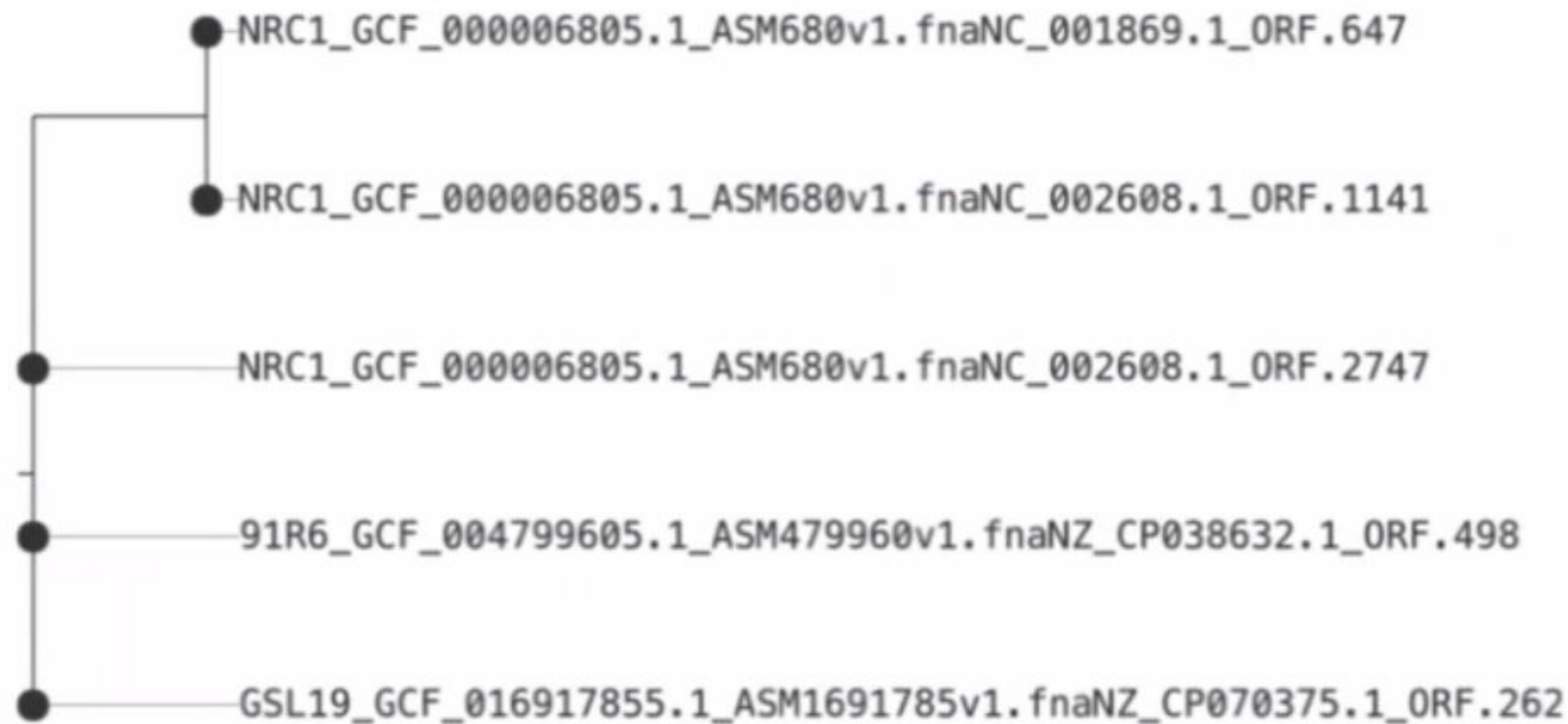


6: Ecoli_GCF_000005845.2_ASM584v2.fna

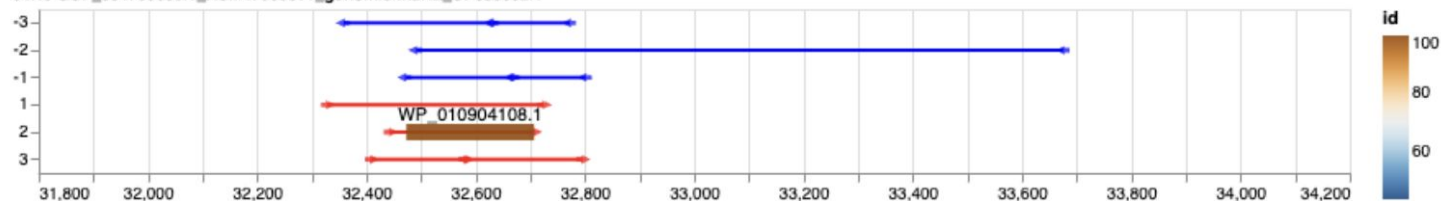


Run Workflow

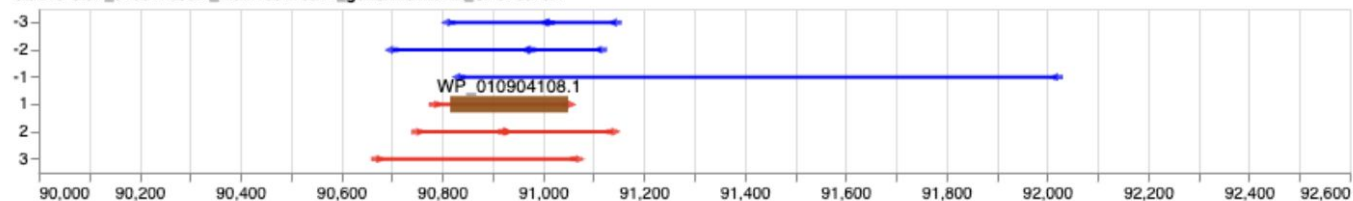




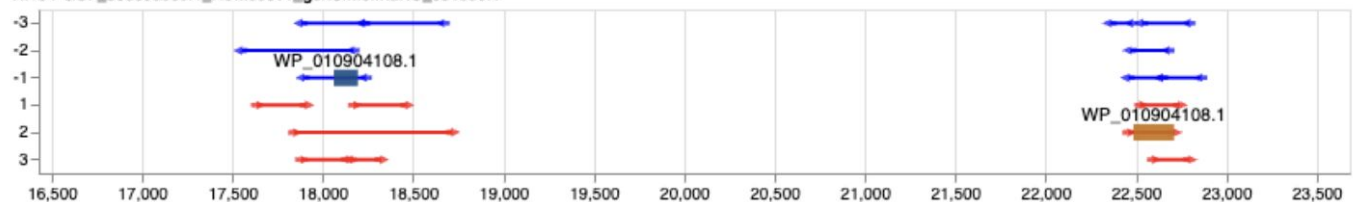
91R6-GCF_004799605.1_ASM479960v1_genomic.fnaNZ_CP038632.1



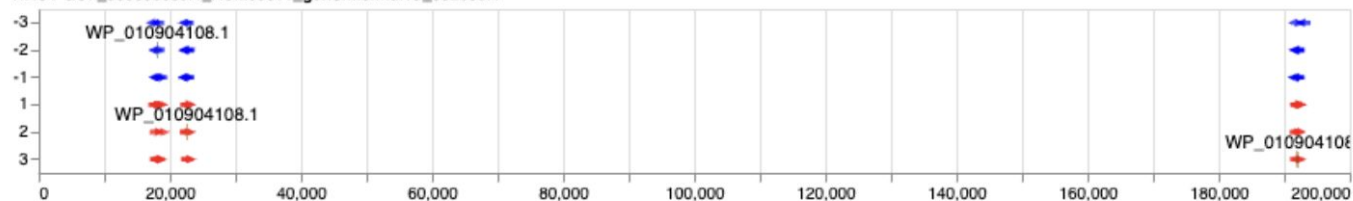
GSL19-GCF_016917855.1_ASM1691785v1_genomic.fnaNZ_CP070375.1



NRC1-GCF_000006805.1_ASM680v1_genomic.fnaNC_001869.1



NRC1-GCF_000006805.1_ASM680v1_genomic.fnaNC_002608.1



Conclusion

Gas vesicle proteins are one of the features that make GSL-19 and other halophiles so special!

GSL-19 is a particularly interesting halophile. It exhibits strong halophilic characteristics such as high GC content/acidity, abundant genes showcasing osmotic balance and membrane adaptations, and (of course) gas vesicle protein genes making it an ideal halophile to study.

Implications

The study of *Halobacterium sp. GSL-19* provides valuable insights into extremophile adaptations.

- **Biotechnology:** enzymes and biopolymers stable in high salt.
- **Astrobiology:** models for extraterrestrial life.
- **Evolutionary Biology:** understanding genome reduction and adaptation mechanisms.

