

# **SNP MATRIX RETRIEVAL**

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# About

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**BACKGROUND**

# Arabidopsis Thaliana

Arabidopsis thaliana is a model plant species widely utilized in scientific research.

Family – **Brassicaceae family** (thale cress)

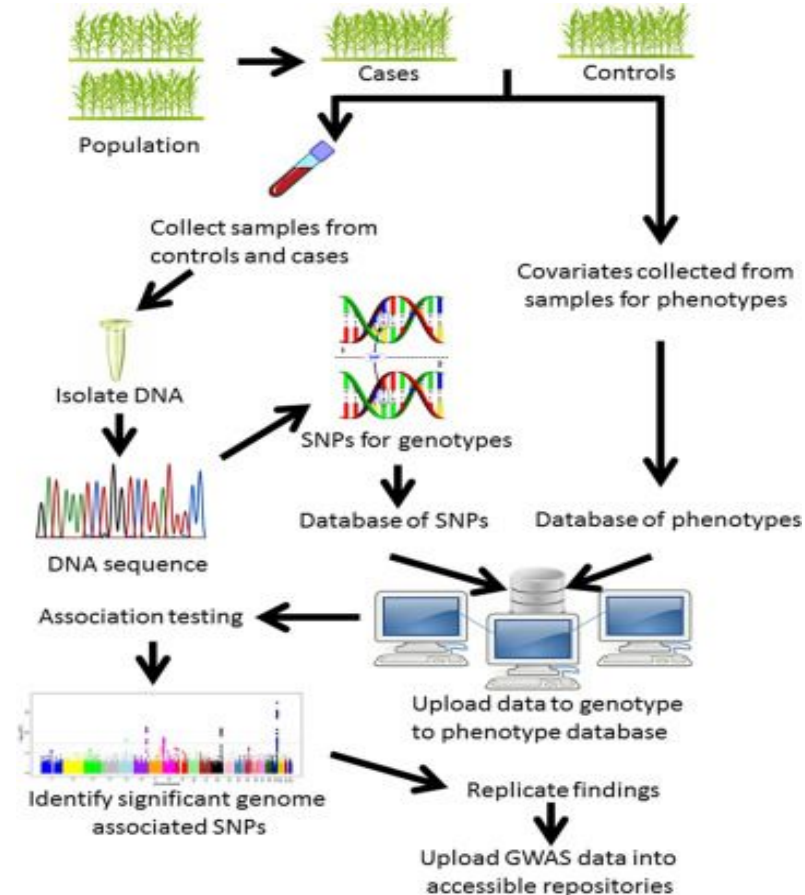
## CHARACTERISTICS:

- **Small Genome Size:** has a relatively small genome compared to other plant species, with approximately 135 million base pairs. This compact genome makes it easier to study and analyze genetic variations.
- **Short Life Cycle:** The life cycle of Arabidopsis thaliana is relatively short, completing its growth and reproduction within a few weeks. This fast life cycle allows researchers to conduct experiments and observe multiple generations in a relatively short period.
- **Genetic Tool:** It possesses a wide array of genetic tools and resources, including a fully sequenced genome, mutant collections, and genetic transformation techniques.



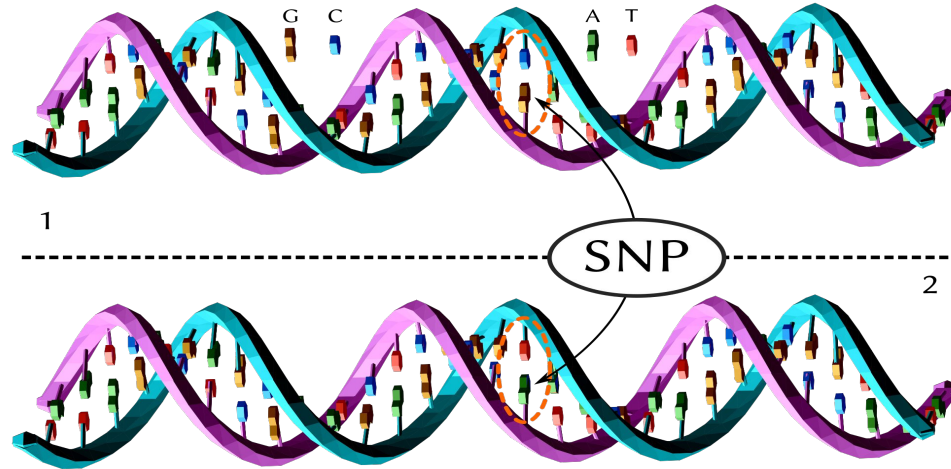
# GWAS

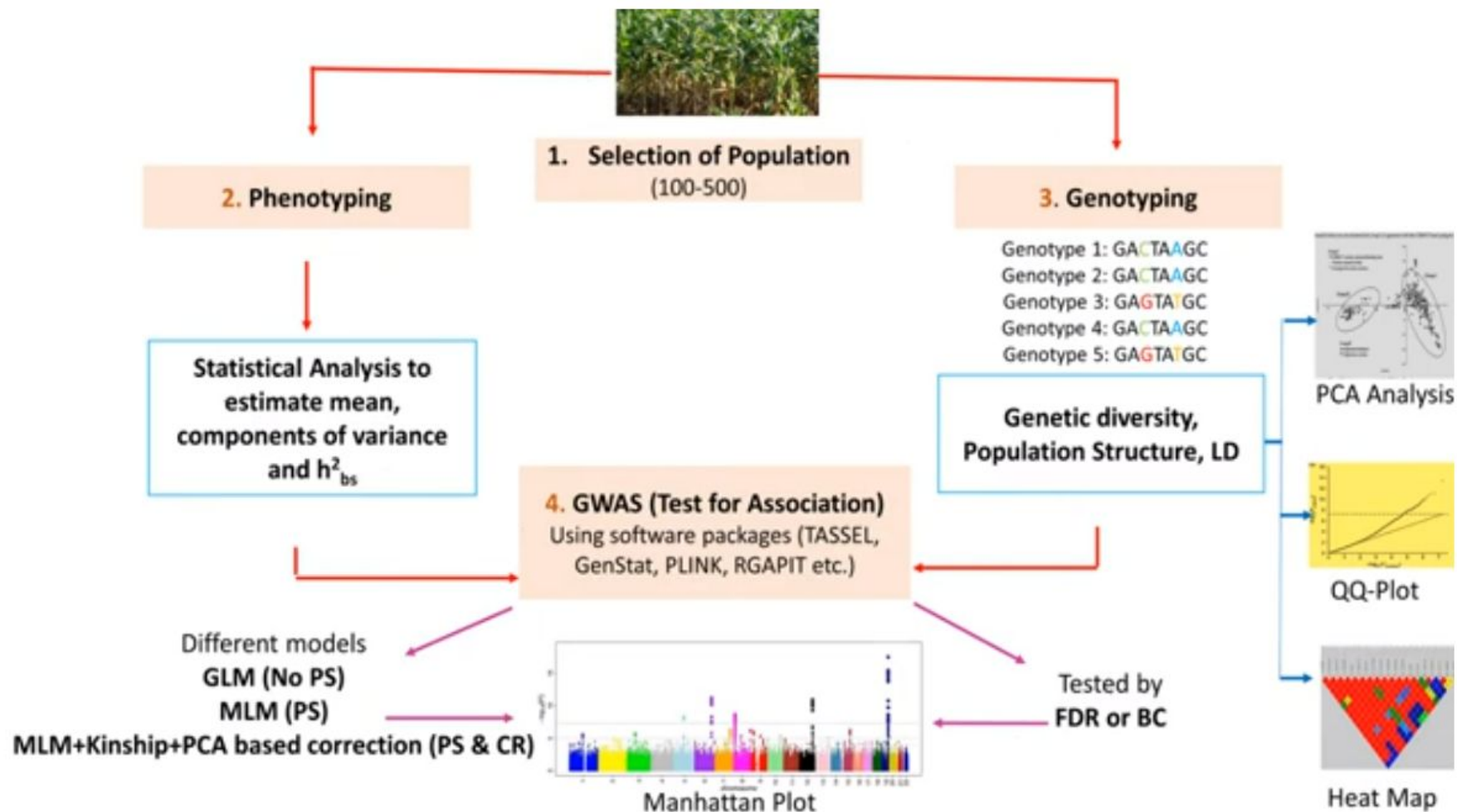
- GWAS analyzes millions of genetic variants across the entire genome.
- It focuses on common genetic variants with modest effects on the trait or disease of interest.
- GWAS reveals the polygenic nature of traits and identifies multiple loci associated with the phenotype.
- It provides insights into the genetic architecture and heritability of traits.
- Integration with functional genomics helps understand the biological mechanisms underlying genetic associations.
- GWAS has advanced our understanding of genetic contributions to complex traits and diseases.



# SNPs

- Genetic marker used in paper.
- Single nucleotide differences in a DNA sequence.
- SNPs present in the coding regions can affect protein function or gene expression ,leading to different phenotypes or traits.





# 1001 Genome project

[News](#) [Data Providers](#) [Accessions](#) [Tools](#)



## 1001 Genomes

A Catalog of *Arabidopsis thaliana* Genetic Variation.

### Tools

Explore the variants. We maintain several tools for data download, visualization, and analysis.

[Go](#)

### Download

Visit the Data Center and download whole sets of SNPs, indels, SVs, and genome sequences.

[Go](#)

### Get Seeds

Seed sets of natural accessions are available for

[Complete set](#)

80 strains (D. Weigel lab, MPI)

195 strains (J. Ecker lab, Salk)

180 strains (M. Nordborg Lab, GMI)

## The 1001 Genomes Plus Vision

The 1001 Genomes Project was launched at the beginning of 2008 to discover detailed whole-genome sequence variation in at least 1001 strains

<https://1001genomes.org/>



# DATASET

The dataset used for the study that investigated the global pattern of polymorphism in *Arabidopsis thaliana* through the analysis of 1,135 genomes is a comprehensive collection of genomic data from diverse populations of *Arabidopsis thaliana*.

Here are some key points about the dataset:

**Sample Size:** The dataset comprises genomic information from 1,135 individual *Arabidopsis thaliana* plants. This large sample size allows for a robust analysis and enhances the representation of genetic variation within the species.

**Geographic Diversity:** The dataset includes genomes from different geographic regions across the globe. By capturing genetic diversity from various populations, the dataset enables the examination of regional patterns of polymorphism and their potential correlation with environmental factors or historical events.

**Genomic Variation Data:** The dataset provides information about genetic variations, such as single nucleotide polymorphisms (SNPs), insertions, deletions, and structural variations, across the genomes of the 1,135 *Arabidopsis thaliana* plants. These variations serve as the basis for analyzing the global pattern of polymorphism in the species.

# DATASET OVERVIEW

Dataset was in the HDF5 format (Hierarchical Data Format 5 ,is a data file format commonly used for storing and organizing large and complex datasets.) → **132 GB**

The keys in the dataset:

- “Accessions” → shape (2029 ,)
- “Positions” → shape (10709466 , )  
Contains two attributes → [chromosome , chromosomal regions]

Chromosomes = [1 , 2 , 3 , 4 , 5]

- “SNPS”  
Snp → shape (10709466 , 2029)



SNP.ipynb ☆

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```
import h5py, numpy
import pandas as pd
import numpy as np

f = h5py.File('/content/drive/MyDrive/GENOTYPES/4.hdf5', 'r')

# Get all SNP positions for all chromosomes (len=10709949)
positions = f['positions'][:]

# Array of tuples with start/stop indices for each chromosome
chr_regions = f['positions'].attrs['chr_regions']

# Array of SNP positions for all chromosomes, each chromosome is a hash
# with "Chr<N>" as key, and a numpy.array of positions as value.
snp_pos_on_chrs = [
    { "label": "Chr1", "chr_idx": 0, "positions": positions[chr_regions[0][0]:chr_regions[0][1]] },
    { "label": "Chr2", "chr_idx": 1, "positions": positions[chr_regions[1][0]:chr_regions[1][1]] },
    { "label": "Chr3", "chr_idx": 2, "positions": positions[chr_regions[2][0]:chr_regions[2][1]] },
    { "label": "Chr4", "chr_idx": 3, "positions": positions[chr_regions[3][0]:chr_regions[3][1]] },
    { "label": "Chr5", "chr_idx": 4, "positions": positions[chr_regions[4][0]:chr_regions[4][1]] }
]

# Print header
```



SNP.ipynb ☆

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```
[ ] f['positions'].attrs['chr_regions']
```

```
array([[      0, 2597735],
       [2597735, 4466530],
       [4466530, 6660782],
       [6660782, 8427786],
       [8427786, 10709466]])
```

```
[ ] f["accessions"]
```

```
<HDF5 dataset "accessions": shape (2029,), type "|S6">
```

```
[ ] f["snps"]
```

```
<HDF5 dataset "snps": shape (10709466, 2029), type "|i1">
```

```
[ ] f.keys()
```

```
<KeysViewHDF5 ['accessions', 'positions', 'snps']>
```

```
[ ] x=list(f['positions'][:100])
```

```
[ ] y=list(f["accessions"])
```



```
# Loop over all chromosomes
for chr in snp_pos_on_chrs:

    # Loop over all positions
    for pos in np.nditer(chr["positions"]):
        # Find index of a specific position
        ix = np.where(chr["positions"] == pos)[0][0]

        # Add chromosome start position to SNP position
        ix = ix + chr_regions[chr["chr_idx"]][0]

        # Get the corresponding SNPs for that position
        snps = f['snps'][ix]

        # Count 0s in snps
        cnt_zeros = np.count_nonzero(snps==0)

        # Count 1s in snp
        cnt_ones = np.count_nonzero(snps==1)

    print(chr["label"], pos, cnt_zeros, cnt_ones
          , ", ".join(snps.astype(str)), sep=",")
```



# Libraries/Packages

- H5py
- Numpy
- Pandas

# VCF (Variant Call Format) Approach



```
In [7]: callset['variants/REF']
```

```
Out[7]: array(['C', 'C', 'C', ..., 'C', 'T', 'T'], dtype=object)
```

```
In [8]: DF = allel.vcf_to_dataframe('/home/newuser/Downloads/7000.vcf')
```

```
In [22]: DF
```

```
Out[22]:
```

	CHROM	POS	ID	REF	ALT_1	ALT_2	ALT_3	QUAL	FILTER	PASS
0	1	85	.	C	NaN	NaN	NaN	25.0		True
1	1	86	.	C	NaN	NaN	NaN	25.0		True
2	1	87	.	T	NaN	NaN	NaN	25.0		True
3	1	90	.	A	NaN	NaN	NaN	25.0		True
4	1	106	.	A	NaN	NaN	NaN	28.0		True
...	...	...	...	...	...	...	...	...	...	...
16508	5	26975398	.	T	NaN	NaN	NaN	32.0		True
16509	5	26975399	.	A	NaN	NaN	NaN	32.0		True
16510	5	26975400	.	G	NaN	NaN	NaN	32.0		True
16511	5	26975401	.	G	NaN	NaN	NaN	32.0		True
16512	5	26975401	.	G	NaN	NaN	NaN	32.0		True

100548737 rows x 9 columns

```
In [9]: DF = allel.vcf_to_dataframe('/home/newuser/Downloads/7000.vcf', fields = ['POS', 'REF', 'ALT_1'])  
/home/newuser/.local/lib/python3.10/site-packages/allel/io/vcf_read.py:1240: UserWarning: 'ALT_1' INFO head  
und  
warnings.warn('%r INFO header not found' % name)  
/home/newuser/.local/lib/python3.10/site-packages/allel/io/vcf_read.py:1454: UserWarning: no type for field  
s/ALT_1', assuming object  
warnings.warn('no type for field %r, assuming %s' % (f, normed_types[f]))  
/home/newuser/.local/lib/python3.10/site-packages/allel/io/vcf_read.py:1564: UserWarning: no number for field  
nts/ALT_1', assuming 1  
warnings.warn('no number for field %r, assuming 1' % f)
```

```
In [24]: DF
```

```
Out[24]:
```

	POS	REF	ALT_1
0	85	C	NaN
1	86	C	NaN

```
allel/opt/io_vcf_read.pyx in allel.opt.io_vcf_read.vcf_skip_variant()
allel/opt/io_vcf_read.pyx in allel.opt.io_vcf_read.FileInputStream.advance()
allel/opt/io_vcf_read.pyx in allel.opt.io_vcf_read.FileInputStream._bufferup()
OSError: [Errno 5] Input/output error
```

In [ ]: DF

```
In [10]: lis=list(DF["POS"][:100])
print(lis[:100])
```

```
[85, 86, 87, 90, 106, 107, 108, 109, 110, 111, 112, 113, 114, 115, 116, 117, 118, 119, 120, 121, 122, 123, 124, 125,
126, 127, 128, 129, 130, 131, 135, 136, 137, 138, 139, 140, 141, 142, 143, 144, 145, 146, 147, 148, 149, 150, 151, 1
52, 153, 154, 155, 156, 157, 158, 159, 160, 161, 162, 163, 164, 165, 166, 167, 168, 169, 170, 171, 172, 173, 174, 17
5, 176, 177, 178, 179, 180, 181, 182, 183, 184, 185, 186, 187, 188, 189, 190, 191, 192, 193, 194, 195, 196, 197, 19
8, 199, 200, 201, 202, 203, 204]
```

```
In [11]: p=list(DF["REF"])
q=list(DF["ALT_1"])
```

```
In [12]: import h5py, numpy
import pandas as pd
import numpy as np
```

```
f = h5py.File('/home/newuser/Downloads/4.hdf5', 'r')
```

```
In [15]: x=list(f['positions'][:100])
print(x)
```

```
[55, 56, 63, 73, 75, 80, 88, 92, 94, 98, 101, 110, 112, 116, 123, 125, 126, 128, 135, 138, 139, 150, 161, 167, 176,
179, 188, 190, 196, 201, 203, 208, 209, 213, 219, 221, 222, 223, 229, 232, 237, 241, 242, 253, 266, 270, 276, 284, 2
86, 288, 291, 298, 301, 306, 311, 314, 317, 322, 323, 324, 332, 334, 342, 346, 348, 349, 352, 353, 363, 364, 375, 38
6, 390, 391, 395, 396, 405, 419, 422, 425, 431, 432, 434, 442, 465, 471, 479, 481, 484, 497, 502, 508, 524, 528, 53
0, 540, 541, 542, 544, 548]
```

In [ ]:

```
In [16]: y=[]
# //store position in common
z=[]
# //store atgc
print(len(lis))
for i in range(len(lis)):
    if lis[i] in x:
```

In [ ]:

In [16]:

```
y=[]  
# //store position in common  
z=[]  
# //store atgc  
print(len(lis))  
for i in range(len(lis)):  
    if lis[i] in x:  
        y.append(lis[i])  
        if (q[i]!='A' and q[i]!='T' and q[i]!='G' and q[i]!='C'):  
            z.append(p[i])  
        else:  
            z.append(q[i])  
    # print(p[i],q[i],z)  
print(z)  
  
100  
['G', 'A', 'G', 'C', 'G', 'G', 'T', 'C', 'T', 'G', 'G', 'C', 'T', 'G', 'T', 'G', 'G', 'T', 'A', 'A']
```

In [17]: print(len(z))

20

In [ ]:

SAMSUNG

**SAMSUNG**

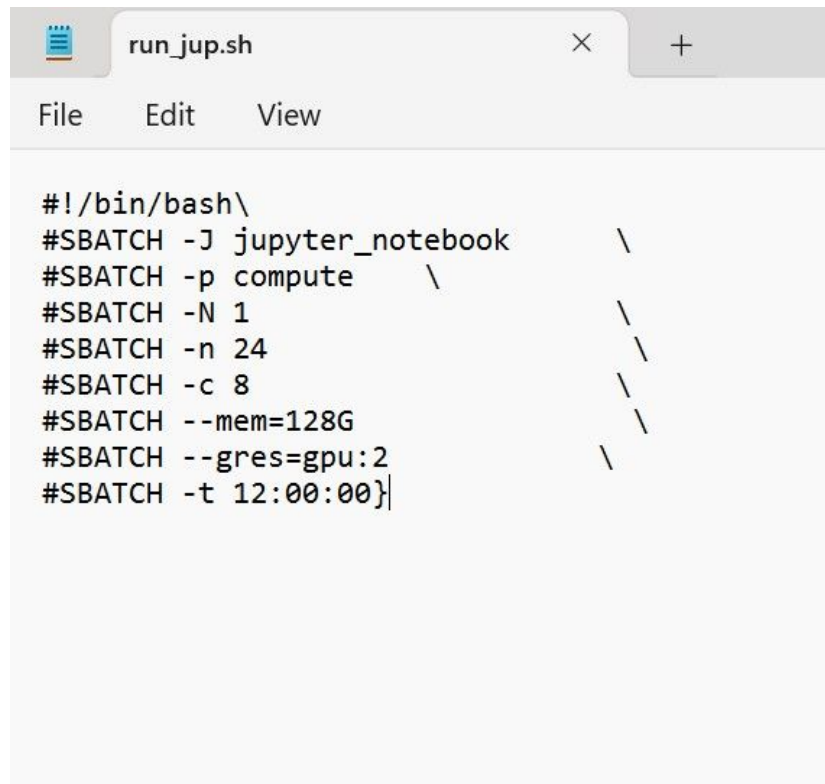
HPC SERVER

## Work Done on HPC –

- ❑ Learning to work on HPC using slurm
- ❑ Setting up suitable environment to run file on slurm
- ❑ Setting up Jupyter notebook as per requirement of the project to run on slurm
- ❑ Testing the environment



# Creating S-batch File According to our needs.



```
#!/bin/bash\  
#SBATCH -J jupyter_notebook \  
#SBATCH -p compute \  
#SBATCH -N 1 \  
#SBATCH -n 24 \  
#SBATCH -c 8 \  
#SBATCH --mem=128G \  
#SBATCH --gres=gpu:2 \  
#SBATCH -t 12:00:00}
```

**Contributions:-**



# Ayush

- Written code for Extracting Data from VCF file as per desired format .
- Worked on HPC server and learned to use slurm
- Setting up Jupyter Environment and solving errors while doing so.
- Created Demonstration using small Dataset of how the code on actual Data will work.
- Analyzing and Detecting the right Approach when discussing with my Professors. Giving updates to Professor.
- Knowing about the Dataset in hands to work with it and code accordingly