

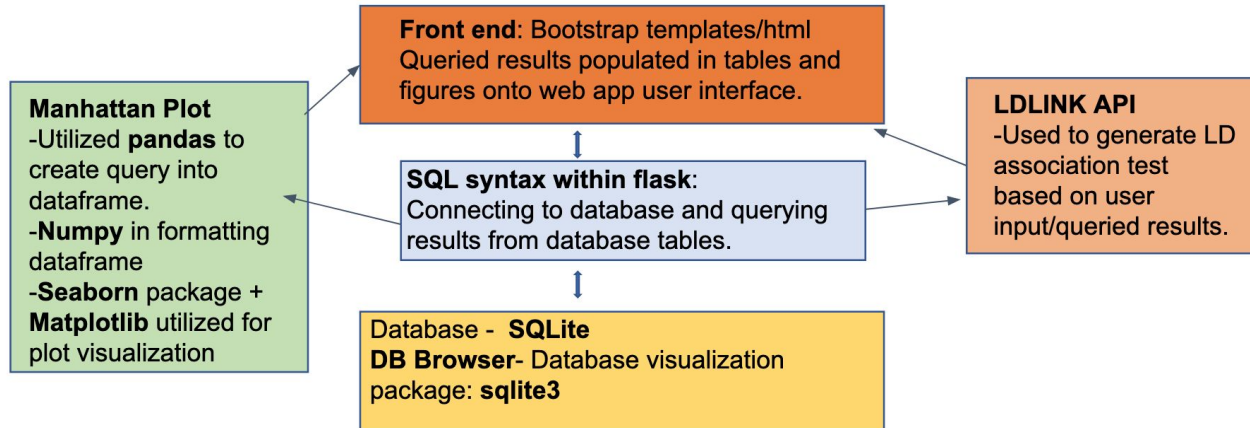


# Type I Diabetes Web Application Presentation

Team Google Chrome



# Software Architecture

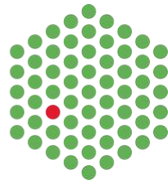




## Data Sources

❏ GWAS data

EMBL-EBI



❏ Gene Ontology data



❏ Population Allele Frequencies and CADD data

SNPnexus





# Connecting the Database with Flask

```
#connecting to database every time we generate a new route
con = db.connect("GC.db", check_same_thread=False)
cursor = con.cursor() #this allows us to query our database
#this is selecting specific info on snp
snp_name = snp_name.lower()
#queried data is executed below to get all info from tables
cursor.execute ("""SELECT gwas.snp, gwas.Gene_name,gwas.p_value,population.Chromosome,
population.Position, population.REF_Allele,
population.ALT_Allele, population.Minor_Allele, population.AFR_Frequency,
population.AMR_Frequency, population.EAS_Frequency, population.EUR_Frequency,
population.SAS_Frequency, CADD.Raw_Score, CADD.PHRED

FROM gwas
INNER JOIN CADD ON gwas.snp = CADD.snp
INNER JOIN population on CADD.snp = population.snp
WHERE gwas.snp= '%s' """ % snp_name)
#^^^^Very important to include '%s' because that is substituted with snp_name^

search_snp = cursor.fetchall()
```

1. Connection
2. Cursor for Query
3. Execution of Query
  - a. 'SELECT' - Columns
  - b. 'INNER JOIN' - Selects tables with matching values
  - c. WHERE - filters the results based on user input.

## Flask redirecting/routing



# Flask

web development,  
one drop at a time

```
#This is the home page
@app.route('/', methods = ['GET', 'POST'])
def index():
    form = QueryForm()
    snp_name = None
    if form.validate_on_submit():
        snp_name = form.snp_name.data
        #Redirect user to snp page if rs is typed
        if snp_name[:2] == "rs":
            return redirect(url_for('SNP', snp_name= snp_name))
        #Redirect user to chromosome page if chr is typed
        if snp_name[:3] == "chr":
            return redirect(url_for('Chromosome',snp_name=snp_name))

        #Redirect user to region page if they enter two locations separated by a comma
        elif "," in snp_name:
            return redirect(url_for('Region', snp_name=snp_name))

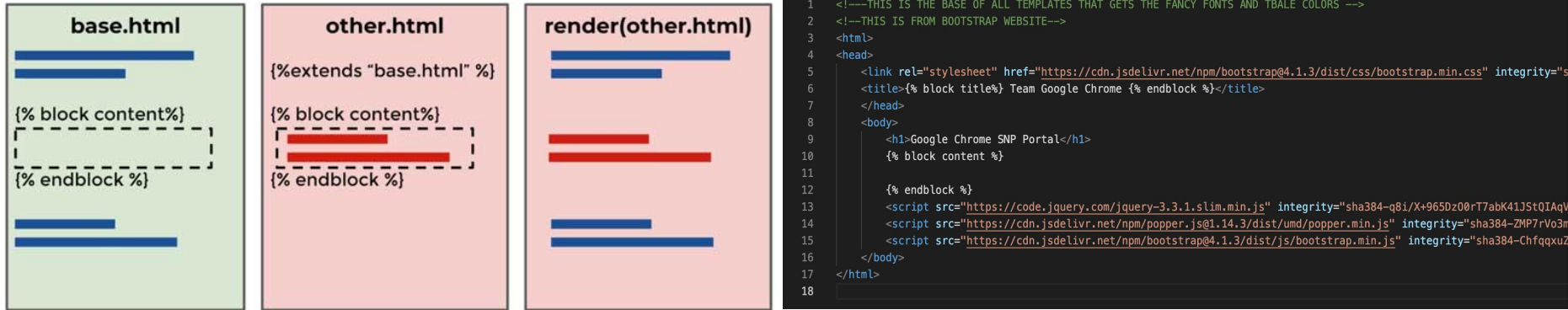
        #redirect to mapped gene page if the beginning != rs or chr or ,
        elif snp_name != 'rs' or snp_name != 'chr' or "," not in snp_name:
            return redirect(url_for('MAPPED_GENE', snp_name=snp_name))

        #redirects to the "SNP" url down below app route
        return redirect(url_for('SNP', snp_name= snp_name))

    return render_template("index.html", form = form, snp_name=snp_name)
```

- Redirect to route based on user input
- Mapped genes unique so if user input does not = to any statement above then redirect to 'MAPPED\_GENE'
- Comma included to separate two positions.

# HTML template inheritance



## Limitations:

Limited to only just chromosome 6 information

## Future development:

Allow clinician to filter chromosome at top of tables.

Condensed tables for gene ontology terms.

## Linkage Disequilibrium - retrieving data



```
for pair in itertools.combinations(rsid_list, r=2):
    rsid1 = pair[0]
    rsid2 = pair[1]
    server = "https://ldlink.nci.nih.gov/LDlinkRest/ldpop?"
    #inputs rsid and population into url for retrieval
    ext =
'var1={rsid1}&var2={rsid2}&pop=ALL&r2_d=r2&genome_build=grch38&token=b56c4bea4225'.
format(rsid1=rsid1, rsid2=rsid2)
```





## LD - Pros and Cons of LDlink API

### Pros

- Apply to whole genome easily

### Cons


- Efficiency - long queries take time
- API blocking
- Token-based access



# Linkage Disequilibrium - User queries



## Linkage Disequilibrium Analysis

Select population of interest:  

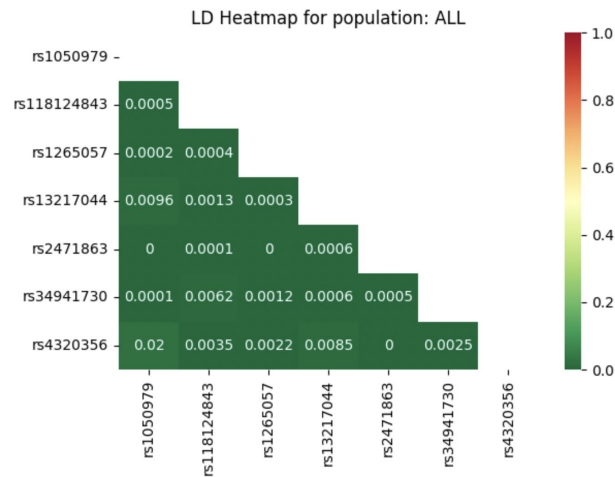
Select your SNPs (larger queries will take longer): \*

rs1050979  
rs9405661  
rs13217044  
rs12203596  
rs4320356  
rs114631266  
rs2471863  
rs9260151  
rs34941730  
rs886424

# Linkage Disequilibrium - displaying and downloading

| SNP Pair                      | r <sup>2</sup> Value |
|-------------------------------|----------------------|
| ('rs1050979', 'rs13217044')   | 0.0096               |
| ('rs1050979', 'rs4320356')    | 0.0197               |
| ('rs1050979', 'rs2471863')    | 0.0                  |
| ('rs1050979', 'rs34941730')   | 0.0001               |
| ('rs1050979', 'rs118124843')  | 0.0005               |
| ('rs1050979', 'rs1265057')    | 0.0002               |
| ('rs13217044', 'rs4320356')   | 0.0085               |
| ('rs13217044', 'rs2471863')   | 0.0006               |
| ('rs13217044', 'rs34941730')  | 0.0006               |
| ('rs13217044', 'rs118124843') | 0.0013               |

Heatmap of LD values



seaborn

Download text file here

[Click here to download a text file of the LD values in a matrix](#)

# Manhattan Plot- Retrieving Data

## Creation of dataframe from SQL Query

```
man_info=[]
for row in search_snp:
    rs_ID = row[0]
    p_value = row[2]
    location = row[3]
    new_row=[rs_ID, p_value, location]
    man_info.append(new_row)

# create dictionary with rsid, pvalue and location data; create pandas dataframe
data_dict = {'snp': [row[0] for row in man_info],
             'p_value': [row[1] for row in man_info],
             'location': [row[2] for row in man_info]}

df=pd.DataFrame.from_dict(data_dict)
```

| snp        | p_value      | location  | chr | -log_pv   | i  |
|------------|--------------|-----------|-----|-----------|----|
| rs1050979  | 6.000000e-14 | 410417    | 6   | 13.221849 | 0  |
| rs9405661  | 2.000000e-09 | 424915    | 6   | 8.698970  | 1  |
| rs13217044 | 4.000000e-06 | 8226764   | 6   | 5.397940  | 2  |
| rs12203596 | 6.000000e-06 | 17120009  | 6   | 5.221849  | 3  |
| rs4320356  | 3.000000e-08 | 26423332  | 6   | 7.522879  | 4  |
| ...        | ...          | ...       | ..  | ...       | .. |
| rs6931865  | 4.000000e-06 | 143758717 | 6   | 5.397940  | 60 |
| rs212408   | 1.000000e-15 | 159049210 | 6   | 15.000000 | 61 |
| rs9356171  | 9.000000e-06 | 163922743 | 6   | 5.045757  | 62 |
| rs73043122 | 4.000000e-06 | 166969779 | 6   | 5.397940  | 63 |
| rs924043   | 8.000000e-09 | 170063801 | 6   | 8.096910  | 64 |



# Manhattan Plot- Formatting

To write P-values in E- notation

```
i=0
for p_value_f in df['p_value']:
    df.at[i, 'p_value'] = p_value_f.replace(' x 10', 'e')
    i+=1
```

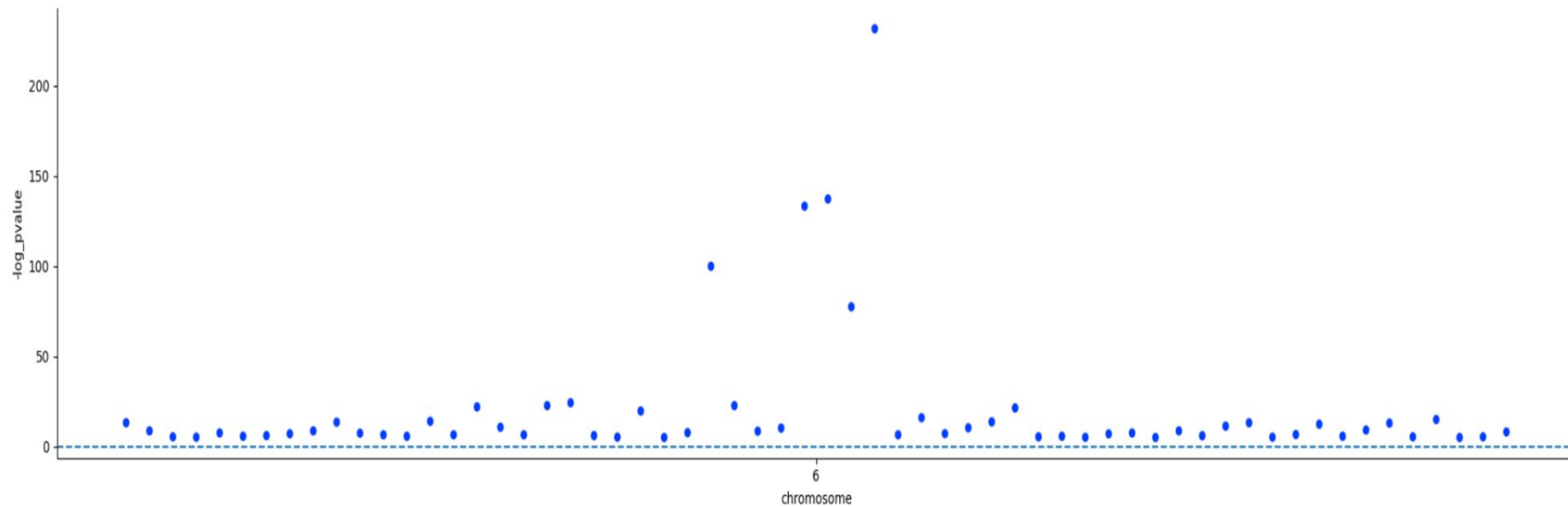
To separate the original location column into two: Chromosome number and location

```
df['chromosome'] = np.vectorize(lambda x: x.split(':')[1])(np.array(df['location'], dtype=str))
df['location'] = np.vectorize(lambda x: x.split(':')[1])(np.array(df['location'], dtype=str))
```



# Manhattan Plot

Manhattan plot showing association between SNPs and Type I Diabetes





# Limitations

- Expand data to other chromosomes
- Make the tables easier to read and filterable
- Make the plots interactive and more informative(e.g. Annotation)
- Improve LD retrieval method

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# Demonstration



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**Any questions?**