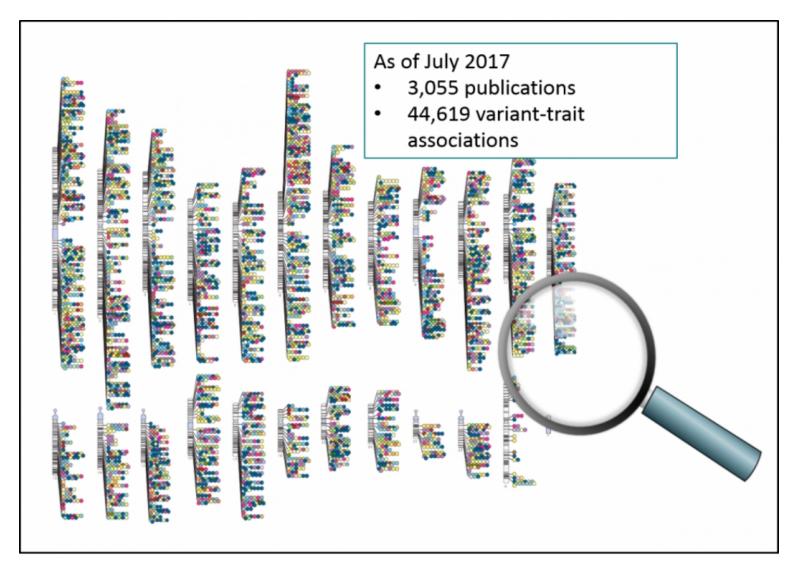
Gene-Gene /SNP-SNP Interaction: BIOFILTER

GBIO0002
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The combinatorial problem of jointly analyzing the millions of genetic variations accessible by high-throughput genotyping technologies is a difficult challenge.

Published in final edited form as:

Pac Symp Biocomput. 2009; : 368-379.

Biofilter: A Knowledge-Integration System for the Multi-Locus Analysis of Genome-Wide Association Studies*

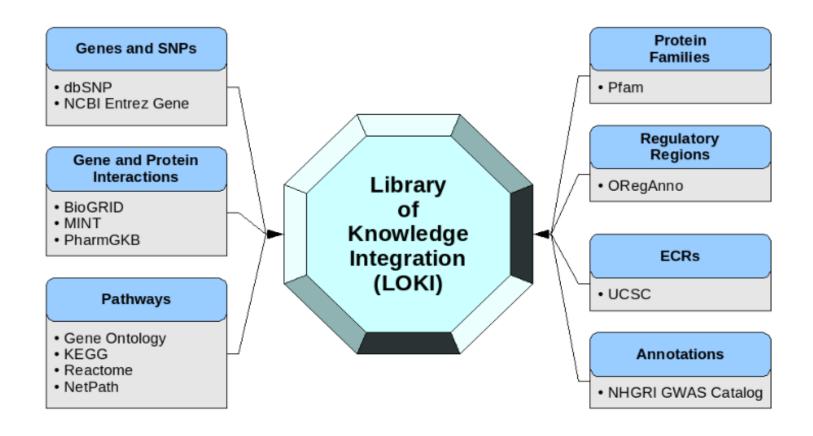
William S. Bush, Scott M. Dudek, and Marylyn D. Ritchie
Center for Human Genetics Research, Vanderbilt University, Nashville, TN 37232, USA

Abstract

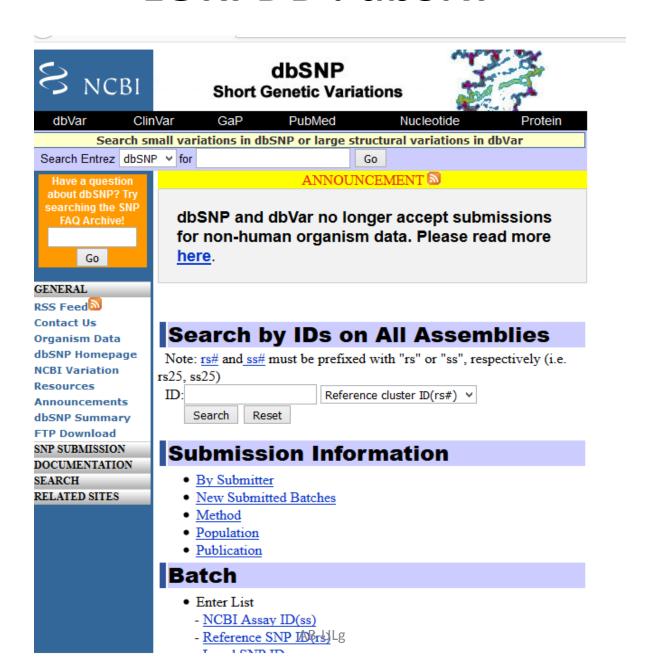
Genome-wide association studies provide an unprecedented opportunity to identify combinations of genetic variants that contribute to disease susceptibility. The combinatorial problem of jointly analyzing the millions of genetic variations accessible by high-throughput genotyping technologies is a difficult challenge. One approach to reducing the search space of this variable selection problem is to assess specific combinations of genetic variations based on prior statistical and biological knowledge. In this work, we provide a systematic approach to integrate multiple public databases of gene groupings and sets of disease-related genes to produce multi-SNP models that have an established biological foundation. This approach yields a collection of models which can be tested statistically in genome-wide data, along with an ordinal quantity describing the number of data sources that support any given model. Using this knowledge-driven approach reduces the computational and statistical burden of large-scale interaction analysis while simultaneously providing a biological foundation for the relevance of any significant statistical result that is found.

☐ Biofilter uses publicly available databases to establish relationships between gene-products

LOKI: Library of Knowledge Integration



LOKI DB: dbSNP



LOKI DB: KEGG database



KEGG PATHWAY Database

Wiring diagrams of molecular interactions, reactions and relations

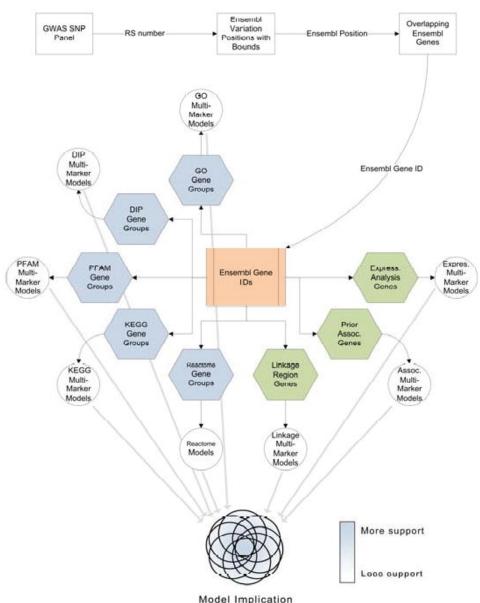
Select prefix map Organism Nsa Organism Organism Nsa Organism Organism Nsa Organism [New pathway maps Update history] REGG PATHWAY is a collection of manually drawn pathway maps representing our knowledge on the molecular interaction, reaction and relation networks for: 1. Metabolism Global/overview Carbohydrate Energy Lipid Nucleotide Amino acid Other amino Glycan Cofactor/vitamin Terpenoid/PK Other secondary metabolite Xenobiotics Chemical structure 2. Genetic Information Processing 3. Environmental Information Processing 4. Cellular Processes 5. Organismal Systems 6. Human Diseases 7. Drug Development KEGG PATHWAY is a reference database for Pathway Mapping. Pathway Identifiers Each pathway map is identified by the combination of 2-4 letter prefix code and 5 digit number (see KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway ko reference pathway highlighting KOs	Menu PATH	WAY BRITE MODULE KO GENES	LIGAND NETWO	RK DISEASE	DRUG DBGET	
REGG PATHWAY is a collection of manually drawn pathway maps representing our knowledge on the molecular interaction, reaction and relation networks for: 1. Metabolism Global/overview Carbohydrate Energy Lipid Nucleotide Amino acid Other amino Glycan Cofactor/vitamin Terpenoid/PK Other secondary metabolite Xenobiotics Chemical structure 2. Genetic Information Processing 3. Environmental Information Processing 4. Cellular Processes 5. Organismal Systems 6. Human Diseases 7. Drug Development KEGG PATHWAY is a reference database for Pathway Mapping. Pathway Identifiers Each pathway map is identified by the combination of 2-4 letter prefix code and 5 digit number (see KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway				Go Help		
molecular interaction, reaction and relation networks for: 1. Metabolism Global/overview Carbohydrate Energy Lipid Nucleotide Amino acid Other amino Glycan Cofactor/vitamin Terpenoid/PK Other secondary metabolite Xenobiotics Chemical structure 2. Genetic Information Processing 3. Environmental Information Processing 4. Cellular Processes 5. Organismal Systems 6. Human Diseases 7. Drug Development KEGG PATHWAY is a reference database for Pathway Mapping. Pathway Identifiers Each pathway map is identified by the combination of 2-4 letter prefix code and 5 digit number (see KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway	Pathway Ma	aps	[New	pathway maps	Update history]	
Global/overview Carbohydrate Energy Lipid Nucleotide Amino acid Other amino Glycan Cofactor/vitamin Terpenoid/PK Other secondary metabolite Xenobiotics Chemical structure 2. Genetic Information Processing 3. Environmental Information Processing 4. Cellular Processes 5. Organismal Systems 6. Human Diseases 7. Drug Development KEGG PATHWAY is a reference database for Pathway Mapping. Pathway Identifiers Each pathway map is identified by the combination of 2-4 letter prefix code and 5 digit number (see KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway		· · · · · · · · · · · · · · · · · · ·		esenting our kno	wledge on the	
Pathway Identifiers Each pathway map is identified by the combination of 2-4 letter prefix code and 5 digit number (see KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway	Global/overview Carbohydrate Energy Lipid Nucleotide Amino acid Other amino Glycan Cofactor/vitamin Terpenoid/PK Other secondary metabolite Xenobiotics Chemical structure 2. Genetic Information Processing 3. Environmental Information Processing 4. Cellular Processes 5. Organismal Systems 6. Human Diseases					
Each pathway map is identified by the combination of 2-4 letter prefix code and 5 digit number (see KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway	KEGG PATHWAY is a reference database for Pathway Mapping.					
KEGG Identifier). The prefix has the following meaning: map manually drawn reference pathway	Pathway Identifiers					
·	-		-	ode and 5 digit n	number (see	
ec reference metabolic pathway highlighting EC numbers rn reference metabolic pathway highlighting reactions <org> organism-specific pathway generated by converting KOs to gene identifiers</org>	ko ec rn	reference pathway highlighting KOs reference metabolic pathway highlightir reference metabolic pathway highlightir	ng reactions	gene identifiers		
and the numbers starting with the following:						

Biofilter: Overview

☐GWAS platform SNPs are mapped to Ensemble gene Ids.

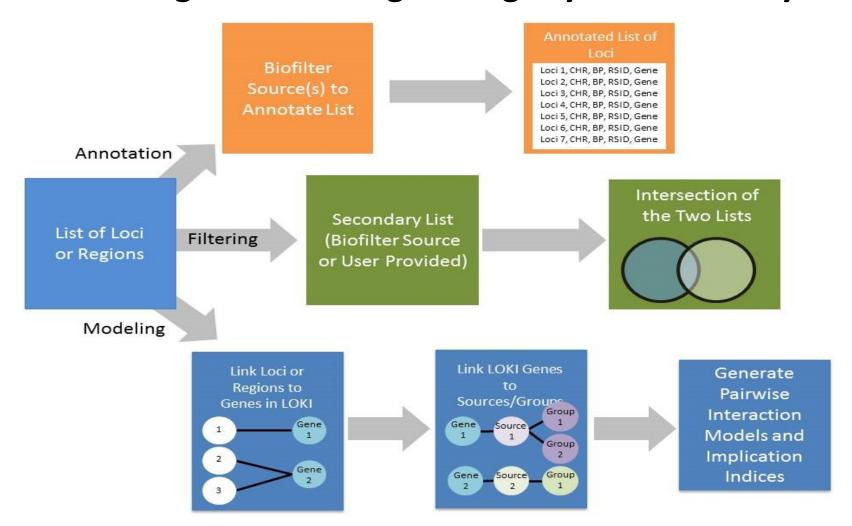
☐Multi-marker models are generated from SNPs within knowledge-related genes.

□Derived models are overlaid to assess overall model implication.



Biofilter: Three Analysis mode

Biofilter has three primary analysis modes and uses the available biological knowledge in slightly different ways.



Biofilter Data types

Data Types

Biofilter can work with and understand the relationships between six basic types of data:

SNP



Specified by an RS number, i.e. "rs1234".

Used to refer to a known and documented SNP whose position can be retrieved from the knowledge database.

Position



Specified by a chromosome and basepair location, i.e. "chr1:234".

Used to refer to any single genomic location, such as a single nucleotide polymorphism (SNP), single nucleotide variation (SNV), rare variant, or any other position of interest.

Region



Specified by a chromosome and basepair range, i.e. "chr1:234-567".

Used to refer to any genomic region, such as a copy number variation (CNV), insertion/deletion (indel), gene coding region, evolutionarily conserved region (ECR), functional region, regulatory region, or any other region of interest.

Gene



Specified by a name or other identifier, i.e. "A1BG" or "ENSG00000121410". Used to refer to a known and documented gene, whose genomic region and associations with any pathways, interactions or other groups can be retrieved from the knowledge database.

Group



Specified by a name or other identifier, i.e. "lipid metabolic process" or "GO:0006629".

Used to refer to a known and documented pathway, ontological group, protein interaction, protein family, or any other grouping of genes, proteins or genomic regions that was provided by one of the external data sources.

Source



Specified by name, i.e. "GO". Used to refer to a specific external data source.

LD Profiles: GWAS information

□Biofilter and LOKI allow for gene regions to be adjusted by the linkage disequilibrium (LD) patterns in a given population.

□When comparing a known gene region to any other region or position (such as CNVs or SNPs), areas in high LD with a gene can be considered part of the gene, even if the region lies outside of the gene's canonical boundaries.

☐ This step require use of additional tool

Biofilter: Command lines vs Configuration

- ☐ Biofilter can be run from a command-line terminal by executing biofilter.py or python biofilter.py
- ☐ All options can either be provided directly on the command line
 - biofilter.py --option-name
- configuration files could be given as input such as

biofilter.py analysis.config

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Biofilter: Command lines vs Configuration

- □Options on the command line are lower-case, start with two dashes and may contain single dashes to separate words (such as "--snp-file"),
- □while in a configuration file the same option would be in uppercase, contain no dashes and instead use underscores to separate words (i.e. "SNP_FILE").

☐ Many command

line options also have alternative shorthand versions of one or a few letters, such as "-s" for "--snp-file" and "--aag" for "--allow-ambiguous-genes".

Output Options: Mode of analysis

--filter / FILTER

Argument: <type> [type] [...] Default: none

Perform a filtering analysis which outputs the specified type

--annotate / ANNOTATE

Argument: <type> [type] [...] [:] <type> [type] [...] Default: none

--model / MODEL

Argument: <type> [type] [...] [:] [type] [...] Default: none

Server Connectivity

- □Connect to ms801 machine
- ☐ Use your login and password

```
_ 0
                                              ms801.montefiore.ulg.ac.be - PuTTY
bhardwaj@ms801:~/DATA$
bhardwaj@ms801:~/DATA$
bhardwaj@ms801:~/DATA$
bhardwaj@ms801:~/DATA$ ls
ann1.config ann7.config
                                    biofilter.TP53
                                                    mod1.config
                                                                    test11.config
                                                                                   test4.config
                                                                                                  test9.config
ann2.config ann8.config
                                                    mod2.config
                                                                   test12.config test5.config
                                    file.map
ann3.config biofilter.gene-group
                                                                                   test6.config
                                    gene
                                                    snp1
                                                                    test1.config
ann4.config biofilter.gene.models
                                    gene2
                                                                    test2.config
                                                                                   test7.config
                                                    snp2
ann5.config biofilter.log
                                                                    test3.config
                                                                                   test88.config
                                    gene-9
ann6.config biofilter.THSD7A
                                                    test10.config test3.region
                                                                                   test8.config
                                    group2
bhardwaj@ms801:~/DATA$
```

Get the genes statistic of loki.db

□ Open terminal and type
 biofilter.py --knowledge loki.db --report-gene-name-stats yes
 □ This indicates number of genes belongs to each category

#type	names	unique	ambiguo	us
	117857	115238	2619	
entrez_g	gid	81664	81664	0
uniprot_	pid	32983	32668	315
ensembl	_gid	75453	75369	84
pharmgk	kb_gid	26650	26650	0
refseq_g	gid	189201	189201	0
refseq_p	oid	117448	117448	0
ensembl	_pid	41732	41732	0
hgnc_id	41163	41163	0	
mim_id	17130	17130	0	
vega_id	19138	19123	15	
mirbase	_id	1879	1879	0
unigene	gid	29152	27997	1155

Get the groups statistic of loki.db

Open terminal and type

- biofilter.py --knowledge loki.db --report-group-name-stats yes
- This indicates no of entries represented by each group

Htypo	namoc	uniquo	ambiguo	
#type	names	unique	ambiguo	us
biogrid_id	d	371104	371104	0
go_id	44957	44957	0	
ontology	44957	44957	0	
kegg_id	323	323	0	
pathway	2605	2588	17	
netpath_	id	28	28	0
oreganno	23393	23393	0	
pfam_id	16718	16718	0	
proteinfa	mily	32501	32164	337
pharmgkl	b_id	108	108	0
reactome	e_id	2163	2163	0
ucsc_ecr	77858	77858	0	

Comparison of Two SNPs list

- ☐ Download snp1 and snp2 from course website
- ☐ Create file name test1.config

KNOWLEDGE loki.db

SNP_FILE snp1

SNP_FILE snp2

FILTER snp

- ☐ Run biofilter.py test1.config
- □As a result , you will get two output files named as biofilter1.log , biofilter1.snp

biofilter1.log

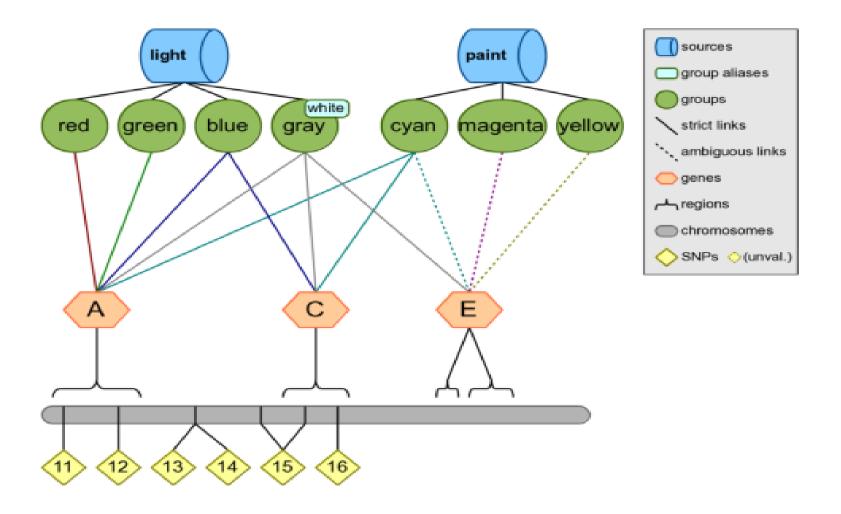
```
loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
adding to main SNP filter ...
... OK: added 4 SNPs (1 RS#s merged)
reducing main SNP filter ...
... OK: kept 1 SNPs (3 dropped, 0 RS#s merged)
writing 'snp' filter to 'biofilter.snp' ...
.. OK: 1 results
```

biofilter1.snp

```
#snp
rs62653571
```

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Let us find the SNPs falling on genes (1)



Let us find the SNPs falling on genes (2)

☐ Create file name test2.config , define snp1 as input

```
KNOWLEDGE loki.db

SNP_FILE snp1

GENE_FILE gene

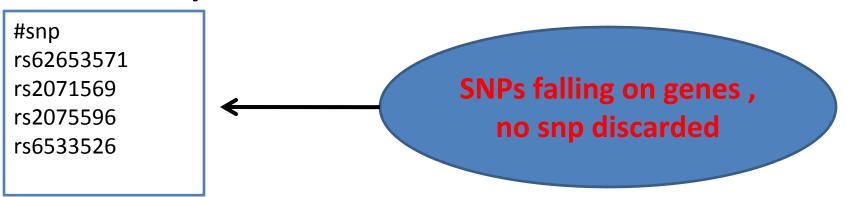
FILTER snp ← Type of filter
```

- ☐Run as biofilter.py test2.config
- ☐ As a result you will get two files: biofilter.log biofilter.snp

biofilter.log

```
loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
adding to main SNP filter ...
... OK: added 4 SNPs (1 RS#s merged)
adding to main gene filter ...
... OK: added 2 genes
writing 'snp' filter to 'biofilter.snp' ...
... OK: 4 results
```

biofilter.snp



Let us find the groups contains specific "regions"

☐ create regions.config which contain group information "R-HSA-5083635"

KNOWLEDGE loki.db GROUP R-HSA-5083635 FILTER region

- ☐Run as biofilter.py regions.config
- ☐As a result you will get two files:

biofilter.log

biofilter.region

biofilter.region : It will consist of all regions belongs to group

#chr	region	start	stop		
X	CFP	47624213	47630305		
15	THBS1	39581079	39598918		
6	THBS2	169215780)	169254114	l .
5	SEMA5A	9035026	9546121		
1	ADAMTS4	161189725	5	161199080)
4	ADAMTS3	72280969	72568799		
5	ADAMTS2	179110851	L	179345430)
21	ADAMTS1	26836287	26845409		
9	ADAMTSL	2	133532164	l .	133575519
4	SPON2	1166932	1208962		
11	SPON1	13962637	14268133		
9	ADAMTS1	3	133414339)	133459403
11	ADAMTS8	130404923	3	130428993	3
21	ADAMTS5	26917912	26967120		
15	ADAMTS7	78759203	78811464		
5	ADAMTS6	65148736	65482014		
13	B3GLCT	31199975	31332276		
7	SSPO	149776042	2	149833965	;

Output a list of all genes within a data source

☐ Create group.config which contain source information

KNOWLEDGE loki.db SOURCE biogrid pfam FILTER gene

- ☐Run as biofilter.py group.config
- ☐ As a result you will get two files: biofilter.log biofilter.gene

biofilter.gene: All genes belong to group PFAM and biogrid

#gene
TRIM54
HDGF
EXOSC10
JMJD6
MC4R
NEDD8

COPS7A

COG1

SIAH1......SO ON

Can you count the number of genes belong to "PFAM" only ????

Let us find genes associated with a pathway or group

□ Create tt.config which contains the Genes detail

KNOWLEDGE loki.db
GENE THSD7A COG8 UBC
FILTER gene snp region group source

- ☐Run as biofilter.py tt.config
- ☐ As a result, you will get two files biofilter.log biofilter.snp.region.group.source

biofilter.log

```
loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
adding to main gene filter ...
... OK: added 3 genes
```

writing 'gene snp region group source' filter to 'biofilter.gene-snp-

region-group-source' ...

... OK: 1668612 result

Open file and check information types

biofilter.snp.region.group.source : It consist of following entries (top 15 lines)

#gene snp chr re	gion	start stop group sou	urce	
THSD7A rs983143041	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs101598290	0 7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs962840243	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs558399301	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs539304291	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs974293917	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs921514204	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs571860735	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs932952501	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs750203807	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs945700395	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs104276294	1 7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs114612380	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs370567942	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs912527472	7	THSD7A 11370435	11832198	biogrid:612143 biogrid
THSD7A rs934667503	7	THSD7A 11370435	11832198	biogrid:612143 biogrid

Let us find a list of genes falling within a group.

☐ Let us create ge-gr.config file which contains the list of genes and group

KNOWLEDGE loki.db GENE HIST1H3A KIAA2013 PQBP1 DCAF8 GROUP R-HSA-5173214 FILTER gene group

☐Run as biofilter.py test.config

□output: biofilter.log, biofilter.gene-group

√biofilter.log

loading knowledge database file '/usr/local/bin/loki.db' ... OK

knowledge database genome build: GRCh38 / UCSC hg38 adding to main gene filter ...

... OK: added 4 genes adding to main group filter ...

... OK: added 1 groups

writing 'gene group' filter to 'biofilter.gene-group' ...

... OK: 0 results



Open file and check information types

Annotating a SNP with gene region information

☐ Let us create annotate.config file which contains the list of snps

KNOWLEDGE loki.db SNP rs11 rs24 rs99 ANNOTATE snp region

☐Run as biofilter.py annotate.config

□output : biofilter.log , biofilter.gene-group

□biofilter.log,

loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
adding to main SNP filter ...
... OK: added 3 SNPs (0 RS#s merged)
writing 'snp : region' annotation to 'biofilter.snp.region' ...
... OK: 3 results

□biofilter.snp-region

Open file and check information types

Annotating SNPs with location information

☐ Let us create annotate2.config file which contains the list of snps

KNOWLEDGE loki.db
SNP rs11 rs24 rs99
ANNOTATE snp position

☐Run as biofilter.py annotate2.config

□output: biofilter.log, biofilter.snp-position

loading knowledge database file '/usr/local/bin/loki.db' OK knowledge database genome build: GRCh38 / UCSC hg38 adding to main SNP filter OK: added 3 SNPs (0 RS#s merged) writing 'snp: position' annotation to 'biofilter.snp.position' OK: 3 results **□**biofilter.snp-region

□ Open file and check chromosomal number where SNPs are present and also define position.
 □ Can you calculate distance among SNPs (in base pairs)

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Map a SNP to the groups and sources where the SNP is present

☐ Let us create annotate3.config file which contains the list of snps

KNOWLEDGE loki.db SNP rs11 rs24 rs99 ANNOTATE snp group source

☐Run as biofilter.py annotate3.config

□output : biofilter.log , biofilter.snp-group-source

biofilter.log

```
database
                                   file
                                         '/usr/local/bin/loki.db'
loading
          knowledge
                                                                 OK
knowledge
                                  build: GRCh38
                                                      UCSC
            database
                       genome
                                                               hg38
adding
                          main
                                        SNP
                                                    filter
               to
       OK:
               added
                          3
                                 SNPs
                                           (0
                                                  RS#s
                                                            merged)
writing 'snp: group source' annotation to 'biofilter.snp.group-source' ...
... OK: 10 results
```

Open file and check information types

Annotating a base pair region with the list of SNPs in that region.

□Let us create annotate4.config file which contains genome position and chromosomal number

KNOWLEDGE loki.db

REGION 1:30000:40000

ANNOTATE snp region

☐Run as biofilter.py annotate4.config

□output : biofilter.log , biofilter.snp-regions

biofilter.log

```
loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
WARNING: UCSC hg# build version was not specified for region input;
assuming it matches the knowledge database
adding to main region filter ...
... OK: added 1 regions
writing 'snp : region' annotation to 'biofilter.snp.region' ...
calculating main region zone coverage ... OK
... OK: 88 results
```

88 SNPs falling in that regions

□biofilter.snp-regions

#snp chr region start stop
rs534702355 1 chr1:30000-40000 30000 40000
rs867282737 1 chr1:30000-40000 30000 40000
rs62028215 1 chr1:30000-40000 30000 40000
rs778316262 1 chr1:30000-40000 30000 40000
rs28688489 1 chr1:30000-40000 30000 40000
rs28628742 1 chr1:30000-40000 30000 40000
rs28594168 1 chr1:30000-40000 30000 40000
rs558169846 1 chr1:30000-40000 30000 40000

..... SO ON

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Count SNPs falling from 6000-10000 genomic regions of chromosome 1, 2, 3, 4 and 5

Create 5 different config files.

Analyse result

Develop the bar graph based on SNPs count.

Mapping regions to genes based on percent of overlap.

□Let us create annotate5.config file which contains genome position and chromosomal number

KNOWLEDGE loki.db

REGION 1:3000:40000

REGION_MATCH_PERCENT 50

FILTER gene

☐Run as biofilter.py annotate5.config

□output : biofilter.log , biofilter.gene

□biofilter.log

loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
WARNING: UCSC hg# build version was not specified for region input;
assuming it matches the knowledge database
adding to main region filter ...
... OK: added 1 regions
writing 'gene' filter to 'biofilter.gene' ...
calculating main region zone coverage ... OK
... OK: 6 results

□biofilter.gene

#gene DDX11L1 MIR1302-2 MIR6859-1 MIR1302-2HG WASH7P FAM138A

Mapping regions to genes based on base pair overlap

□Let us create annotate6.config file which contains genome position and chromosomal number

KNOWLEDGE loki.db REGION 1:4000:10000 REGION_MATCH_BASES 10 FILTER gene

☐Run as biofilter.py annotate6.config

□output : biofilter.log , biofilter.gene

Open biofilter.gene and check genes count

Pair wise Gene-Gene and SNP-SNP interaction

Step 1

Map the input list of SNPs to genes within Biofilter.

Step 2

Connect, pairwise, the genes that contain SNPs in the input list of SNPs.

<u>Step 3</u>

Break down the gene-gene models into all pairwise combinations of SNPs across the genes within sources

Step 1

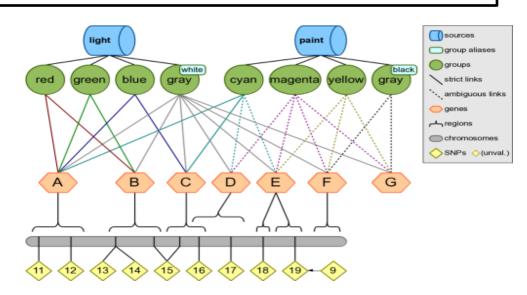
☐ Let us create mod1.config file which contains snp list

KNOWLEDGE loki.db

SNP rs983143041 rs101598290 rs962840243 rs558399301 rs539304291 rs974293917 rs921514204 rs571860735 rs932952501 rs750203807 rs945700395 rs1042762941 rs114612380 FILTER gene

□Run as biofilter.py mod1.config

□output : biofilter.log , biofilter.gene



```
loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
adding to main SNP filter ...
... OK: added 13 SNPs (0 RS#s merged)
writing 'gene' filter to 'biofilter.gene' ...
... OK: 2 results
```

2 SNPs falling in gene regions

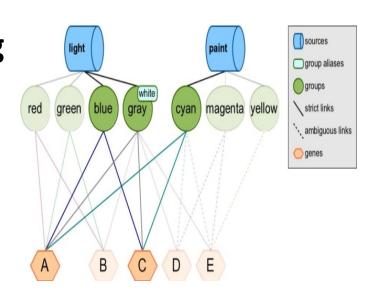
#gene LOC105375153 THSD7A

Step 2

☐ Let us create mod2 .config file which contains gene list observed in step 1

KNOWLEDGE loki.db
GENE LOC105375153 THSD7A
MODEL gene

- ☐Run as biofilter.py mod2.config
- **□**output : biofilter.log , biofilter.gene



biofilter.log

```
loading knowledge database file '/usr/local/bin/loki.db' ...
... OK
knowledge database genome build: GRCh38 / UCSC hg38
adding to main gene filter ...
... OK: added 2 genes
writing 'gene' models to 'biofilter.gene.models' ...
identifying main model candidiates ... OK: 2 candidates
identifying candidiate model groups ... OK: 161848 groups
calculating baseline models ... OK: 0 models
... OK: 0 results
```

O genes validated in modelling step. These genes have no interaction.

☐ Let us create mod11.config file which contains snp list

KNOWLEDGE loki.db

SNP rs268 rs316 rs326 rs328 rs333 rs334 rs544 rs551 rs567 rs662 rs669 rs671 rs683 rs684 rs688 rs689 rs690 rs693 rs694 rs695 rs696 rs698 rs699 rs700 rs703 rs705 rs712 rs715 rs835 rs868 rs900 rs910 rs958 rs1124 rs1164 rs1182 rs1183 rs1208 rs1303 rs1321 rs1421 rs1442 rs1506 rs1510 rs1545 rs1547 rs1590 rs1748 rs2506 rs2566 rs2688 rs2689 rs2765 rs2767 rs2942 rs2962 FILTER gene

☐ Let us create mod22.config file which contains gene list from step 1.

□ Check output . Did you get any output ? YES

score(src-grp) #gene1 gene2 LIPC LPL 4-12 INS **INSR** 3-16 3-6 **APOB** LDLR **APOB** LPL 3-5 3-4 ADH1C ALDH2 **APOB** LIPC 3-4 **LDLR** LIPC 3-4 3-4 **MKKS** CCT5 **SNRPN** 2-8 **SNURF** 2-3 GH1 INS 2-2 INS PAX6 2-2 INS HNF1B

These genes are interacting in pair wise manner

- □Download SNP from NCBI (like 100 snps), check their gene information.
- ☐ Analyse which genes are interacting and back trace the SNPs based on that interaction result.

- ☐ Analyse their genomic position.
- □ Calculate distance among SNPs.