A quick guide to JSNMF algorithm with mouse brain (RNA+H3K4me3, Paired-Tag) data

Load data

```
addpath('../data')
load('h3k4me3_5k_10k.mat')
num_clu = length(unique(label));
```

Marker genes and housekeeping genes are given by user

```
marker_genes = ["Dnah12"; "Cfap299";'Flt1'; 'Slco1a4'; 'Inpp5d'; 'Hexb'; 'Slc1a3';
'Atp1a2'; 'Slc1a2'; 'Gpc5'; 'Prr5l'; 'Plp1'; 'Rnf220'; 'Lhfp13'; 'Cacng4'; 'Erbb4';
'Kcnc2'; 'Reln'; 'Grin3a'; 'Adarb2'; 'Fam19a2'; 'Ahcyl2'; 'Gm32647'; 'Shisa6';
'Epha6'; 'Galnt14'; 'Hs3st4'; 'Tshz2'; 'Fam19a1'; 'Il1rapl2'; 'Cdh12'; 'Rgs6';
'Lingo2'; 'Prr16'; 'Zfp804b'; 'Pex5l'; 'Gm26883'; 'Cdh18'; 'Gm28928'; 'Galnt16';
'Spag16'; 'Cfap43'; 'Wdr49'; 'Ebf1'; 'Mecom'; 'Ptprb'; 'Tgfbr1'; 'Zfhx3'; 'Apbb1ip';
'Gpc5'; 'Plpp3'; 'Slc1a2'; 'Igsf8'; 'St18'; 'Mog'; 'Mag'; 'St18'; 'Mbp'; 'Vcan';
'Tnr'; '6030443J06Rik'; 'Nxph1'; '6330411D24Rik'; 'Kcnmb2'; 'Grip1'; 'Gm45341';
'Gm45455'; 'Grip1'; 'Trpm3'; 'Rfx3'; 'Dgkh'; 'Hs3st4'; 'Ryr3'; 'Hs6st3'; 'Gm2164';
'Grik4'; 'Cpne7'; 'Gm2164'; 'Ryr3'; 'Hs6st3'; 'Sdk1'; 'Foxp2'; 'Garn13hrm2'; 'Grm8';
'Vwc21'; 'Olfm3'; 'Grik1 Gm2164'; 'Gm28928'; 'Sgcz'; 'Prr16'; 'Chrm3'; 'Pdzrn3';
'Kcnq5'; 'Unc5d'; 'Car10'; 'Pcdh15'; 'Nrg1'];
marker_genes = unique(marker_genes);
load('housekeeping_genes.mat')
housekeeping_genes = hp_genes;
```

Preparing for RAGI

```
load('h3k4me3_rna_RAGI.mat')
genes = genes_h3k4me3; data = rna_h3k4me3;
[genes_h3k4me3, rna_h3k4me3] = rna_ragi(barcodes, share_bd, genes, RNA,1);

[~, ~, ind1] = intersect(marker_genes,genes,'stable');
[~, ~, ind2] = intersect(housekeeping_genes,genes,'stable');
```

Run JSNMF

Parameter selection

```
addpath('../codes')
addpath('../external')
```

```
[alpha, gamma, Inits] = parameter_selection(X1, X2, label);

disp(['value of alpha:',num2str(alpha));
disp(['value of gamma:',num2str(gamma));

value of alpha:0.36157
value of gamma:3734.8227
```

Start time

```
tic
[W1, W2, H1, H2, S, iter, objs] = jsnmf(X1, X2, alpha, gamma, Inits);
disp('JSNMF runtime:');
iteration starts!
number of iteration:10 obj:2167309.0214
number of iteration:20 obj:2160203.2657
number of iteration:30 obj:2157360.4942
number of iteration:40 obj:2155750.9988
number of iteration:50 obj:2154672.8747
number of iteration:60 obj:2153795.968
number of iteration:70 obj:2152994.5733
number of iteration:80 obj:2152529.4533
number of iteration:90 obj:2152224.5205
converged!
JSNMF runtime:
Elapsed time is 89.678800 seconds.
ac: 0.8536 393/2684 nmi:0.8358
ave_marker_gene: 0.4256
                            ave hk gini:0.1896 diff gini:0.2361
```

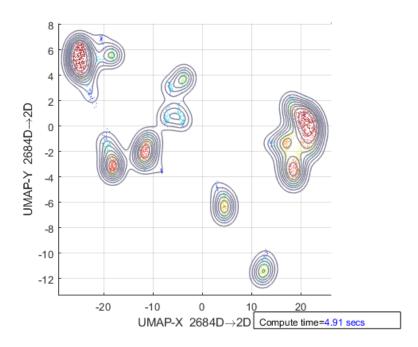
Evaluating performance

```
% KNN on S
A = Wtrim(S,50);
% implement louvain clustering
[clust,~,~] = getNCluster(A,num_clu,0,3,20);
if length(unique(clust))== num_clu
    [ac, nmi_value, ~] = CalcMetrics(label, clust);
    [ave_mk_gini, ave_hk_gini, difgini] = RAGI(data,ind1,ind2,clust);
else
% using spectral clustering instead if number of clusering doesn't equals to the pointed number
    [~, clust, ~] = SpectralClustering(A, num_clu);
    [ac, nmi_value, ~] = CalcMetrics(label, clust);
    [ave_mk_gini, ave_hk_gini, difgini] = RAGI(data,ind1,ind2,clust);
end
```

Downstream analysis including visualization

visualization on S obtained from JSNMF

```
% load UMAP package
addpath('../umapFileExchange/umap')
addpath('../umapFileExchange/util')
term = '.';
label_name = label_h3k4me3(:,2);
label_name = strrep(label_name, '_', '\_');
colors = generateColors(max(length(unique(label name))),length(unique(term)));
D jsnfm = 1-S; D jsnfm = D jsnfm-diag(diag(D jsnfm));
[reduction_jsnmf, ~, ~, ~] =
run umap(D jsnfm, 'metric', 'precomputed', 'min dist', 0.68, 'n neighbors', 12);
gscatter(reduction jsnmf(:,1),reduction jsnmf(:,2),label name,colors,[],4);
set(gca,'xtick',[],'ytick',[]);
title('UMAP for JSNMF')
legend('Location', 'westoutside', 'Box', 'off', 'FontSize', 9.5);
legendmarkeradjust(16)
legend('boxoff')
Parallelizing UMAP with MATLAB's 8 assigned logical cores for nn descent tasks,
sgd tasks
Running basic (ub) reduction, v2.1.3
(ub=basic, us=supervised, ubt=template, ust=supervised template)
Warning: Start point not provided, choosing random start point.
UMAP(method=MEX, n neighbors=12, n components=2, metric='precomputed',
n epochs=[], learning rate=1, init=spectral, min dist=0.68, spread=1,
set op mix ratio=1, local connectivity=1, repulsion strength=1,
negative sample rate=5, transform queue size=4, a=0.341586509870361,
b=1.53964927967393, randomize=true, target n neighbors=-1,
target_metric='categorical', target_metric_kwds=[], target_weight=0.5,
verbose=true, initial alpha=1, sparse data=false, small data=true,
distance func='precomputed', dist args=[]
```



UMAP reduction finished (cost 4.93 secs) Finished basic (ub) reduction

UMAP for JSNMF FC_ExNeu_PT HC_ExNeu_CA1 HC_ExNeu_CA23 FC_ExNeu_L23 HC_ExNeu_DG FC_ExNeu_L5 BR_InNeu_Sst FC_ExNeu_Claustrum FC_ExNeu_L4 BR_InNeu_CGE BR_NonNeu_Microglia BR_InNeu_Pvalb BR_NonNeu_Astro_Myoc FC_ExNeu_CT FC_ExNeu_NP HC_NonNeu_Ependymal BR_NonNeu_Oligo_MOL BR_NonNeu_OPC BR_NonNeu_Endothelial HC_ExNeu_Subiculum

Preparing for gene ontology (GO) enrichment analysis

```
% analysis on W1
W1 = full(W1);
% normalization is not necessary
% W1_norm = W1 * diag(1./sqrt(sum(W1.^2)));
```

```
% sort each column of W1 to find key genes
[B, I] = sort(W1, 1, 'descend');
nfactor = size(W1, 2);
```

loading genes and peaks

```
load('h3k4me3_5k_10k.mat','genes','peaks')

% selecting top 200 genes with large values
topk = 200;
IX = cell(topk+1, 1);
IX{1} = 'gene symbol';

% save top genes for each factor, and these genes can be used for pathway
% enrichment analysis.

% need to construct Directory 'kidney\W1' in advance
for i = 1:nfactor
    filenm = ['H3K4me3\W1\factor_' num2str(i) '.csv'];
    IX(2:end) = genes(I(1:topk, i));
    dlmcell(filenm, IX);
    disp(['Successfully write factor_',num2str(i),' for gene enrichment analysis']);
end
```

```
Successfully write factor_1 for gene enrichment analysis Successfully write factor_2 for gene enrichment analysis Successfully write factor_3 for gene enrichment analysis Successfully write factor_4 for gene enrichment analysis Successfully write factor_5 for gene enrichment analysis Successfully write factor_6 for gene enrichment analysis Successfully write factor_7 for gene enrichment analysis Successfully write factor_8 for gene enrichment analysis Successfully write factor_9 for gene enrichment analysis Successfully write factor_10 for gene enrichment analysis Successfully write factor_11 for gene enrichment analysis Successfully write factor_12 for gene enrichment analysis Successfully write factor_13 for gene enrichment analysis Successfully write factor_14 for gene enrichment analysis Successfully write factor_14 for gene enrichment analysis
```

```
% analysis on W2(region loading)
% check peaks with abnormal name, such as {'chrUn-GL456239-39648-39894'},
% these peaks should be removed from peak list
% peaks(1101,:) = [];
[B,I] = sort(W2, 1, 'descend');
```

```
nfactor = size(W2,2);
topk = 1000;
IX = cell(topk,nfactor);
loci = cell(topk,3);
% check peak names with special symbols
peaks = strrep(peaks,':','-');
% need to construct Directory 'kidney/W2' in advance
for i = 1:nfactor
   IX(:,i) = peaks(I(1:topk,i),:);
   IX(:,i) = regexp(IX(:,i), '-', 'split');
   filenm = ['H3K4me3\W2\factor_' num2str(i) '.bed'];
   for j = 1:topk
       loci{j, 1} = char(IX{j,i}{1,1});
       loci{j, 2} = char(IX{j,i}{1,2});
       loci{j, 3} = char(IX{j,i}{1,3});
       dlmcell(filenm, loci);
   disp(['Successfully write factor_',num2str(i),' for region enrichment analysis']);
end
% after obtaining genes from W1 and peaks from W2, gene ontology (GO) enrichment
% can be implemented using some online tools, such as GREAT, Metascape and
```

```
Successfully write factor_1 for region enrichment analysis Successfully write factor_2 for region enrichment analysis Successfully write factor_3 for region enrichment analysis Successfully write factor_4 for region enrichment analysis Successfully write factor_5 for region enrichment analysis Successfully write factor_6 for region enrichment analysis Successfully write factor_7 for region enrichment analysis Successfully write factor_8 for region enrichment analysis Successfully write factor_9 for region enrichment analysis Successfully write factor_10 for region enrichment analysis Successfully write factor_11 for region enrichment analysis Successfully write factor_12 for region enrichment analysis Successfully write factor_13 for region enrichment analysis Successfully write factor_14 for region enrichment analysis Successfully write factor_14 for region enrichment analysis
```