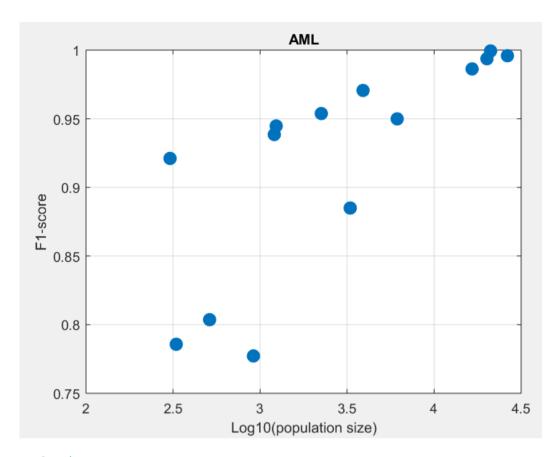
## **AML Dataset Classification**

```
%% Read the Data and Preprocess
>> DataTable = readtable('AML benchmark.csv');
% remove unneeded columns
>> DataTable.Time=[];
>> DataTable.Cell length=[];
>> DataTable.DNA1=[];
>> DataTable.DNA2=[];
>> DataTable.Viability=[];
>> DataTable.file number=[];
>> DataTable.event number=[];
>> DataTable.subject=[];
% Separate Data points and Labels
>> Labels=DataTable.cell type;
>> DataTable.cell type=[];
>> Data = table2array(DataTable);
>> clear DataTable
% clear NotDebrisSinglets
>> Data(strcmp('NotDebrisSinglets', Labels),:)=[];
>> Labels(strcmp('NotDebrisSinglets', Labels))=[];
% Apply arcsinh5 transformation
>> Data=asinh((Data-1)/5);
%% run LDA Classifier with 5-fold cross-validation
>> CVO = cvpartition(Labels, 'k', 5);
>> Accuracy = zeros(CVO.NumTestSets,1);
>> training time = zeros(CVO.NumTestSets,1);
>> testing time = zeros(CVO.NumTestSets,1);
>> CellTypes = unique(Labels);
>> ConfusionMat = zeros(length(CellTypes));
>> for i = 1:CVO.NumTestSets
>>
       trIdx = CVO.training(i);
       teIdx = CVO.test(i);
>>
>>
       tic
>>
       classificationLDA = fitcdiscr(...
           Data(trIdx,:), ...
>>
           Labels(trIdx));
>>
>>
      training time(i)=toc;
                                      %in seconds
>>
>>
>>
       Predictor = predict(classificationLDA, Data(teIdx,:));
>>
      testing time(i)=toc;
                                      %in seconds
>>
       Accuracy(i) =nnz(strcmp(Predictor,Labels(teIdx))) ...
       /size(Labels(teIdx),1);
       ConfusionMat = ConfusionMat + ...
>>
       confusionmat(Labels(teIdx), Predictor, 'order', CellTypes);
```

```
>> end
>> Total time = sum(training time)+sum(testing time);
>> training time = mean(training time);
>> testing time = mean(testing time);
>> cvAcc = mean(Accuracy) *100;
>> cvSTD = std(Accuracy) *100;
>> disp(['LDA Accuracy = ' num2str(cvAcc) ' ' char(177) ' '...
   num2str(cvSTD) ' %'])
LDA Accuracy = 98.1437 \pm 0.018051 \%
>> clear i Predictor classificationLDA trIdx teIdx CVO Accuracy
%% Performance evaluation
% F1 measure
>> Precision = diag(ConfusionMat)./sum(ConfusionMat,1)';
>> Recall = diag(ConfusionMat)./sum(ConfusionMat, 2);
>> Fmeasure = 2 * (Precision.*Recall)./(Precision+Recall);
>> MedianFmeasure = median(Fmeasure);
>> Subset size = sum(ConfusionMat,2);
>> WeightedFmeasure = (Subset size./size(Data,1))'*Fmeasure;
>> disp(['Median F1-score = ' num2str(MedianFmeasure)])
Median F1-score = 0.94711
>> figure, scatter(log10(Subset size), Fmeasure, 100, 'filled')
>> title('AML')
>> xlabel('Log10(population size)'), ylabel('F1-score')
>> box on, grid on
```

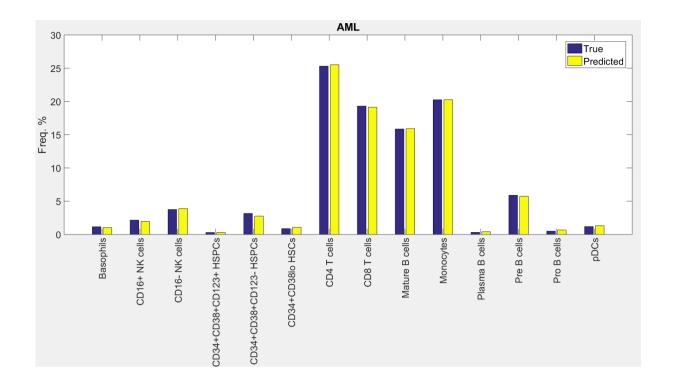


```
%% Population Frequency
>> True_Freq = sum(ConfusionMat,2)./sum(sum(ConfusionMat));
>> Predicted_Freq = sum(ConfusionMat,1)'./sum(sum(ConfusionMat));
>> Max_Freq_diff = max(abs(True_Freq-Predicted_Freq))*100;

>> disp(['delta_f = ' num2str(Max_Freq_diff)])

delta_f = 0.40121

>> figure,bar([True_Freq*100 Predicted_Freq*100])
>> xticklabels(CellTypes)
>> xtickangle(90)
>> set(gca,'FontSize',15)
>> legend({'True','Predicted'},'FontSize',15)
>> legend show
>> ylabel('Freq. %'),title('AML')
```



## %% Population Frequency scatter plot

```
>> X=log(True_Freq*100);
>> Y=log(Predicted_Freq*100);
>> figure,scatter(X,Y,50,'filled')
>> box on, grid on
>> xlabel('Log(True frequency %)')
>> ylabel('Log(Predicted frequency %)')
>> title('AML')
>> for k=1:length(CellTypes)
>> text(X(k),Y(k),CellTypes{k})
>> end
>> lsline
>> text(0,0,['R = ' num2str(corr(X,Y))])
```

