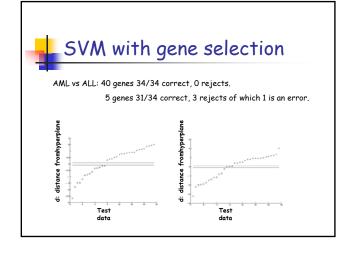




Feature/gene selection

- SVMs as covered in class use all features or genes.
- However, molecular biologists and oncologists believe that only a small number of genes are responsible for particular biological properties.
- When the number of samples is much much smaller than the number of features, over-fitting is very likely. Possible performance improvement (reduction in over-fitting) with fewer features.
- Feature selection is thus a very important problem in classification studies of gene expression data.





Two approaches

- Filter: make an independent assessment based on general characteristics of the data. The feature set is filtered to produce the most promising subset before learning.
- Wrapper: to evaluate the feature subset using the machine learning algorithm that will ultimately be employed for learning. The learning algorithm is wrapped into the feature selection procedure.



Filter technique: Fisher scores

For each feature, compute the fisher index

Fisher(i) =
$$\frac{\mu_{i}^{+} - \mu_{i}^{-}}{(\sigma_{i}^{+})^{2} + (\sigma_{i}^{-})^{2}}$$

 μ_i^+ is the mean of the i^{th} feature among the + class σ_i^+ is the std. dev. of the i^{th} feature among the + class



Feature i



Wrapper technique: Recursive feature elimination

- Solve the SVM problem and find vector w.
- Rank order elements of w by absolute value
- Discard features/genes corresponding to the bottom 10% of the values.
- Retrain SVM on reduced gene set and go back to step 2



Example

Dataset	Sample s	Class 0	Class 1
Leukemia	38	27	11
Morphology (train)		ALL	AML
Leukemia	34	20	14
Morpholgy (test)		ALL	AML
Leukemia Lineage	23	15	8
(ALL)		B-Cell	T-Cell
Lymphoma Outcome (AMI)	15	8 Low risk	7 High risk

	Dataset	Total Sample s	Class 0	Class 1
	Lymphoma Morphology	77	19 FSC	58 DLCL
	Lymphoma Outcome	58	22 Low risk	36 High risk
_	Brain Morphology	41	14 Glioma	27 MD
۲	Brain Outcome	50	38 Low risk	12 High risk

Hierarchy of difficulty:

- $1. \quad \hbox{Histological differences: normal vs. malignant, skin vs. brain} \\$
- 2. Morphologies: different leukemia types, ALL vs. AML
- 3. Lineage B-Cell vs. T-Cell, folicular vs. large B-cell lymphoma
- 4. Outcome: treatment outcome, elapse, or drug sensitivity



Results: part 1

Dataset	Algorithm	Total Samples	Total error s	Class 1 errors	Class 0 errors	Number Genes
Leukemia Morphology (trest)	SVM	35	0/35	0/21	0/14	40
AML vs ALL	WV	35	2/35	1/21	1/14	50
	k-NN	35	3/35	1/21	2/14	10
Leukemia Lineage (ALL) B vs T	SVM	23	0/23	0/15	0/8	10
	WV	23	0/23	0/15	0/8	9
	k-NN	23	0/23	0/15	0/8	10
Lymphoma FS vs DI CI	SVM	77	4/77	2/32	2/35	200
_	WV	77	6/77	1/32	5/35	30
	k-NN	77	3/77	1/32	2/35	250
Brain MD vs Glioma	SVM	41	1/41	1/27	0/14	100
	WV	41	1/41	1/27	0/14	3
	k-NN	41	0/41	0/27	0/14	5

