

# Scientific Computing Semester Review

## 1. MATLAB Basic Command

- 2D plot, 3D plot
- m-file function
- array, matrix manipulation (vector calculation in MATLAB)

## 2. Financial Computing

- Compounding: basic formula:  $f = p(1+r)^n$  / rule of 70 / rule of 200 / how to compute mortgage
- Internal Rate of Return: The rate at an investment plan breaks even

$$NPV = \sum_{i=0}^n \frac{C_i}{(1+r)^i} = 0 \quad NFV = \sum_{i=0}^n C_i(1+r)^{n-i} = 0$$

## 3. K-means clustering

- Objective: Deduction for unlabeled data, difference between input data set and cluster center should be as small as possible.
- Objective function:

$$J(X_{d \times n}; C_{d \times m, m < n}, A_{n \times m}) = \sum_{j=1}^m \sum_{i=1}^n a_{ij} \|x_i - c_j\|^2 \quad A: \text{assignment matrix}$$

- Step: fixed C to find best A, fixed A to find best C, repeat above til converge

$$a_{ij} = \begin{cases} 1 & \text{if } j = \arg \min \|x_i - c_q\|^2 \\ 0, & \text{otherwise} \end{cases} \quad c_i = \frac{\sum_{i=1}^n a_{ij} x_i}{\sum_{i=1}^n a_{ij}}$$

\* guarantee to converge! (J is reduced repeatedly)

## 4. Image Compression

- True color: Each pixel is represented by [R,G,B] vector  
Index-color: Each pixel is represented by an index into a color map of  $2^b$  colors
- Compression Ratio

$$\rho = \frac{\text{before}}{\text{after}} = \frac{m * n * 3 * 8}{m * n * \log_2 c + c * 3 * 8} \approx \frac{24}{\log_2 c}$$

- Blocked Base Compression Ratio(application of k-means clustering):

$$\rho = \frac{\text{before}}{\text{after}} = \frac{m * n * 3 * 8}{\frac{m}{q} * \frac{n}{q} * \log_2 c + c * 3 * 8} \approx \frac{24q^2}{\log_2 c + \frac{24cq^4}{m * n}}$$

## 5. MLE for PDF

- Maximum Likelihood Estimation: A way to find statistical model with optimum data  
 $\Rightarrow$  Formulate objective function and derive parameter  
 $\Rightarrow$  Inequality of Arithmetic and Geometric Means

$$\frac{\sum_{i=1}^n x_i}{n} \geq \left( \prod_{i=1}^n x_i \right)^{1/n}$$

equality holds only when  $x_1 = x_2 = \dots = x_n$

- Probability Density Function:

$$- J(p, q, r) = p^{n1} q^{n2} r^{n3}, \text{ with } p + q + r = 1, p, q, r \geq 0$$

$$- \text{Common- Gaussian Distribution: } g(x, \mu, \sigma^2) = \frac{1}{\sigma\sqrt{2\pi}} \exp[-0.5(\frac{x - \mu}{\sigma})^2]$$

$$\text{MLE: } \mu = \frac{1}{n} \sum x_i \quad \sigma^2 = \frac{1}{n} \sum (x_i - \mu)^2$$

## 6. ML for Classification

- K-nearest Neighbour Classifier: Find first k nearest neighbour of a given point, determine the class by voting among those points.

– Voronoi Diagram

– PROS: intuitive/no computation for model construction

CONS: massive computation when data is big/no straightforward way

- Quadratic Classifiers: Select a class of parameterized PDF then identify parameters via MLE

$$\Rightarrow \text{test stage: } C = \arg \max Pr(C) * pdf_c(x)$$

– Characteristic: decision boundary is quadratic function

– PROS: easy compute when data is small/easy for LOOCV

CONS: covariance matrix is hard to compute or not exist/cannot handle bi-model data

- Naive Bayes Classifiers:

$$\text{Identify class pdf } pdf_c = pdf_{1,c} * pdf_{2,c} * \dots * pdf_{d,c}$$

$$\Rightarrow \text{test stage: } C = \arg \max Pr(C) * pdf_c(x)$$

- Characteristic: statistical independent/each feature governed by a PDF
- PROS:Fast computation/Robust than QC
- CONS:not able to deal bi-model/class boundary not complex as QC
- Deep Neural Network: given desired i/o pairs and construct a model with it(structure identification & parameter identification)
  - ⇒ Multilayer perceptron
  - >3 layers(deep neural network) for arbitrary regions
  - Training: gradient descent(check:sec 11.)/Gauss-Newton method
  - Chain rule & partial derivative to implement a basic math model
- Performance Evaluation: recognition rate↑,error rate↓,computation load↓
  - Stage:training set/validation set/test set
  - Method:inside test/one-side holdout test/two-side holdout test/M-fold cross validation/leave one-out cross validation

## 7. Data Fitting and Regression Analysis

- Data fitting :
  1. Using \ to solve basic regression problem  $A\theta = b \Rightarrow \theta = A \backslash b$
  2. Error sum function is the objective function we want to minimize, can be shown as sum of squared error  $\sum [y_i - f(x_i)]^2$
  3. MATLAB provide `polyfit` and `polyval` to simplify our codes.
- Least-squares Estimate
  1.  $E(\theta) = (A\theta - b)^T(A\theta - b) \nabla_{\theta}(\theta) = 0 \Rightarrow \theta = (A^T A)^{-1} A^T b$
  2. Application: distance to a line/plane
 
$$d^2 = \left( \frac{a}{\sqrt{a^2 + b^2}} x_0 + \frac{b}{\sqrt{a^2 + b^2}} y_0 - \frac{c}{\sqrt{a^2 + b^2}} \right)^2$$
- Non-linear Regression Problem  $\Rightarrow$  Transformation example:  $y = ae^{bx} \Rightarrow \ln y = \ln a + bx$

## 8. Principal Component Analysis

- Objective: Reduction of unlabelled data with dimensionality reduction, data variance ↑
  - \* not design for clasification problem
- i/o: zero justified dataset  $\Rightarrow$  unit vector  $\mathbf{u}$  that square sum of dataset's projection onto  $\mathbf{u}$  is maximized

- Application: line/plane fitting, machine learning, face recognition
- PCA for Total Least Squares: minimum projection on  $\mathbf{y}$  means maximum projection on  $\mathbf{x}$ , so  $\mathbf{y}$  is the normal vector

## 9. Page Rank

- Connectivity Matrix(row:in/column:out) $\Rightarrow$  Transition Probability Matrix  
 $A = pA_1 + (1 - p)A_2$       $A_1$ :follow with the link  $A_2$ :jump to random page  
*PageRank*:  $z$  for initial state,  $Az, A^2z \dots A^kz$ , until  $A^{k+1}z = A^kz$
- Compute Pagerank: eigenvector method/powermethod(for large  $n$ )
- Eigenvalue and Characteristic
  1. Row,Column sum with eigenvalue and eigenvector
  2. Eigenvalue Decomposition  $A^k = VD^kV^{-1}$
- Application: Team ranking, university rank, recommendation

## 10. Audio Basic

- Parameter of recording audio file: sampling rate/Bit solution(8/16bits)/no. of channel
- Time domain vs Frequency domain

	Time	Frequency
features	Period, Intensity, Timbre(Waveform)	Energy, Pitch, Timbre

## 11. Optimization

- Gradient Descent:  $x_{next} = x_{now} - \eta \nabla f(x_{now})$       $\eta$ : learning rate  
 \* not guarantee for global optimum
- DSS

## 12. Dynamic Programming

An effective method for finding optimum solution to a multi-stage decision problem.

1. Characteristic: Decomposition/Subproblem optimality
2. Steps: Optimum-value function  $\rightarrow$  Recurrent formula  $\rightarrow$  Answer
3. Applications: LCS, Edit distance, matrix chain products, hidden Markov model...

## 4. Example :QBSH

Find the alignment between the pitch vector of human's singing and the note vector of a given melody, both in the unit of semitone or MIDI number.

$$dist(p, q) = \min_{u_1, u_2, \dots, u_m} \sum_{i=1}^m |p(i) - q(u_i)|,$$

subject to  $u_1 = 1$  and  $u_1 \leq u_2 \leq u_3 \cdots \leq u_m$ .