

## From last meetings

Class Web Page:

<http://www.stat.unc.edu/faculty/marron/321FDAhome.html>

Finding Structure in Populations of Complex Objects: PCA

Important duality:

Object Space



Feature Space

Cornea Data: motivated robust PCA

# Robust PCA: Toy Example I

E.g. previous “random parabolas”, with an outlier(?) added

Show CurvDat\Parabs1outRaw.ps

## Notes:

- this is **not** a coordinate-wise outlier
- but recall **many** other directions in  $\mathfrak{R}^d$
- very different in “shape” (nearly orthogonal?)
- random parabolas “live in a special part” of  $\mathfrak{R}^d$

## Robust PCA: Toy Example I (cont.)

### Effects of outlier on PCA:

Show CurvDat\Parabs1outCurvDat.ps and again show CurvDat\ParabsCurvDat.ps (flip back and forth)

- Very minor effect on mean (since it feels outliers less)
- Not large effect on PC1 (large variab'ty in that dir'n "wins")

Again show CorneaRobust\OutliersPCA.ps

- Major effect on PC2 (see both proj'ns and mean  $\pm$  ext.)
- Still some effect on PC3
- Major redist'n of Sums of Squares (signal power)
- PC didn't find "best" directions? (but 3d subspace is right)

## Naïve Robust PCA:

### Spearman Correlation:

Idea: base PCA on correlation matrix computed on ranks  
from: “Rank based” nonparametric statistics

(for  $X_1, \dots, X_n$ , ranks are  $(1), \dots, (n)$ , where  $X_{(1)} \leq X_{(2)} \leq \dots \leq X_{(n)}$ )

Result: Small improvements, but outlier is still PC2

Show CurvDat\Parabs1outCurvDatSCorr.ps

# Better Robust PCA

Recall Huber's  $L^1$  M-estimate of "center"

show CorneaRobust\L1Center.ps

Corresponding PCA: work with data projected to sphere

Show CorneaRobust\OutliersPCA.ps and CorneaRobust\ SphericalPCA.ps

Toy Example 1:

Show CurvDat\Parabs1outCurvDatSph.ps and compare with CurvDat\ParabsCurvDat.ps

- PC1 and PC2 very similar to original (except scale)
- Outlier goes into PC3

## Robust PCA: Toy Example II

E.g. previous “random parabolas”, with 2 outliers added

Show CurvDat\Parabs2outRaw.ps

### Ordinary PCA:

Show CurvDat\Parabs2outCurvDat.ps

- PC1 & PC2: feels 1<sup>st</sup> outlier as before
- PC3 & PC4: “tilt” and 2<sup>nd</sup> outlier are confounded
- found right 4d subspace, but poor directions within
- can see outliers in jitter plots / smoothed histograms

## Robust PCA: Toy Example II

Spearman PCA: fails again (essent'ly same as ordinary PCA)

Show CurvDat\Parabs2outCurvDatSCorr.ps

Spherical PCA:

Show CurvDat\Parabs2outCurvDatSph.ps

- PC1 & PC2: similar to no outlier case
- PC3 & PC4: outliers appear here
- Outliers are “mixed” between 3 & 4 (by previous dir'ns)
- i.e. didn't find “nicest 4 directions”

## Numerical Aside

In these toy examples,  $n = 50$ ,  $d = 10$ , so could also have used:

1. Projection pursuit robust PCA
2. PCA based on standard robust covariance matrices

But for the cornea data,  $n = 43$ ,  $d = 66$ , so these don't work.

Thus Spherical PCA is only choice



# Elliptical PCA

Result: Spherical PCA is good, not great, for the cornea data

Idea: problem is “Fourier type signal compression”

“high frequency terms”  $\ll$  “low frequency terms”

Solution: Replace “sphere” by “ellipse”,  
Which reflects “proper scaling (of coordinate axes)”

Problem: simple and computable ellipse?

## Elliptical PCA (cont.)

Three Step solution (keying on “parallel to coordinate axes”):

1. Rescale coordinate axes by Median Absolute Deviation:

$$MAD = \underset{i}{\text{med}} \left| X_i - \underset{i'}{\text{med}} X_{i'} \right|$$

2. Project onto circle
3. Return axes to original scale

# Elliptical PCA for Cornea Data

Show CorneaRobust\NORMLWR.MPG

## PC1:

Show CorneaRobust\NORM100.MPG and CorneaRobust\NORM122.MPG

- robust center slightly better
- same main lesson (about this direction)
- robust slightly better at edge

# Elliptical PCA for Cornea Data

## PC2:

Show CorneaRobust\NORM200.MPG and CorneaRobust\NORM222.MPG

- same main lesson
- outlier edge effect eliminated

## PC3:

Show CorneaRobust\NORM300.MPG and CorneaRobust\NORM322.MPG

- outlier effect eliminated
- main effect slightly diminished (no free lunch)

# Elliptical PCA

Problem: nonlinear analysis leads to **systematic bias**

Approach (N. Locantore): iterative improvement

Cornea Data: not a major issue

## Cornea Data – Radius of analysis

Idea: can control radius of disk

General purpose: 4.2 mm, e.g. study PRK recovery

Show CorneaRobust\EgPRKmorph.mpg

Avoid edge effects (usually completely): 3 mm

Here: 4.0 mm (accentuated impact of robustness!)

# Big Picture

Goal 1: Understanding Population Structure

PCA: illustrated with Cornea Data

Goal 2: Discrimination (classification)

Corpora Callosa data

F. L. D. failed

Now derive “Orthogonal Subspace Projection”