

Using Unsupervised Learning with ICA to Identify Patterns of Glaucomatous Visual Field Defects

M. Goldbaum, et al.
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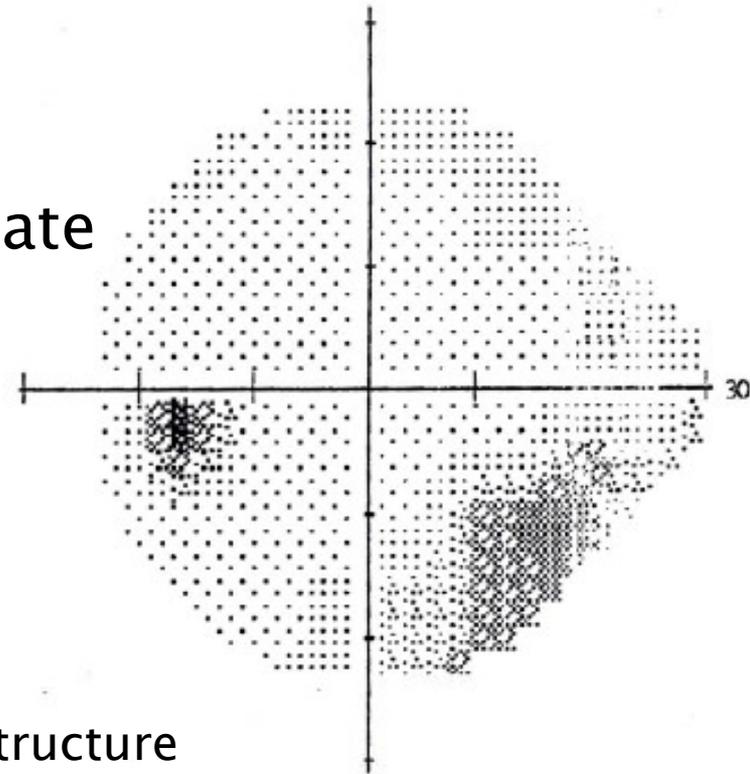


Visual field test

- Prime indicator for functional defects due to glaucoma
 - Detection of progression
 - Diagnosis

- Visual field defect **patterns** indicate glaucoma
 - Expert knowledge
 - Grown over “generations”

- Application of machine learning
 - Classification by supervised learning
 - ▪ Clustering by **unsupervised** learning
 - Organize input data to meaningful structure



Assumption -> Goal of the study

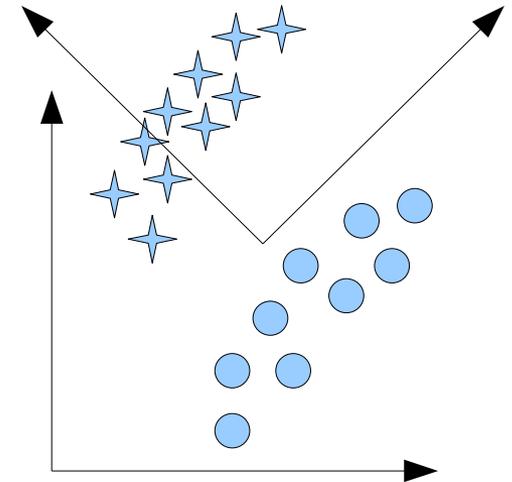


Unsupervised learning techniques
can explore
relevant and meaningful, but hidden patterns
from
(visual field defect) data !



Variational Bayesian ICA mixture model

- **Component Analysis:**
 - Project data onto axes that meaningful represent the data
- **Principal Component Analysis:**
 - Projection maximizes data variance
 - Dimension reduction
- **Independent Component Analysis:**
 - Produce axes that are maximally independent
 - Components are statistically independent
- **One single model may represent data suboptimal**





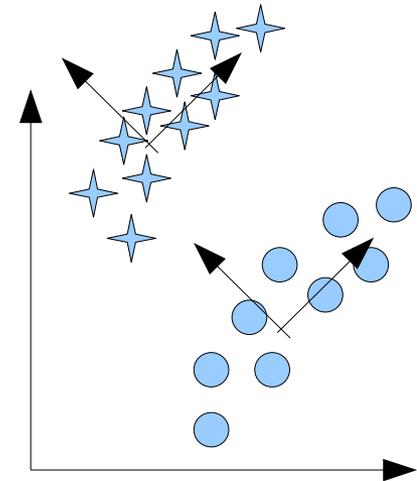
Variational Bayesian ICA mixture model

■ ICA mixture model

- Non linear ICA technique
- Learning multiple ICA models for each cluster
- Weight each instance probabilistically

- Optimizes
 - Axes
 - Number of clusters

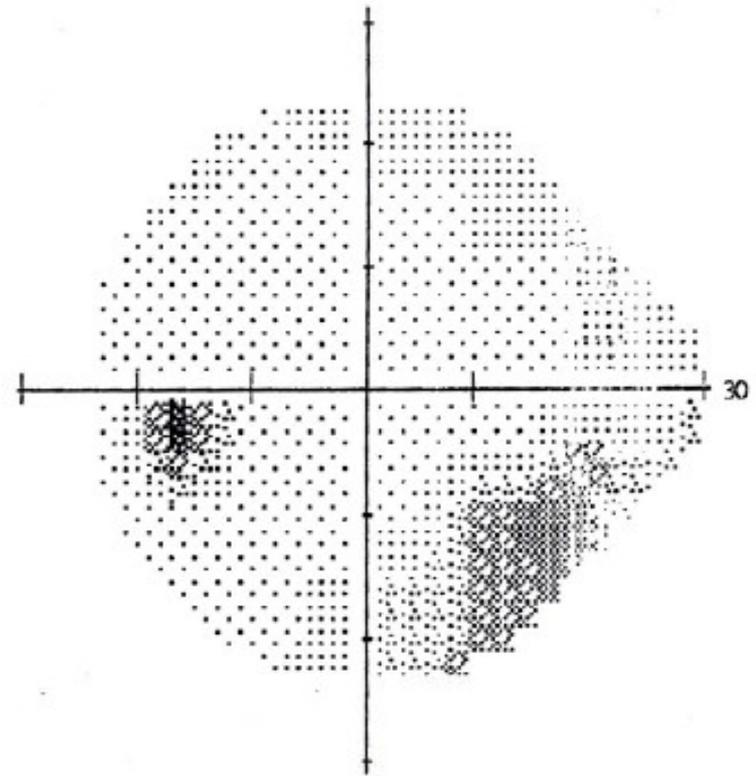
- Optimization via variational Bayesian approx. techniques





Data

- Visual field test
- 345 eyes
 - 156 GON
 - 189 normal
- Feature vector
 - 52 visual field locations
 - Age





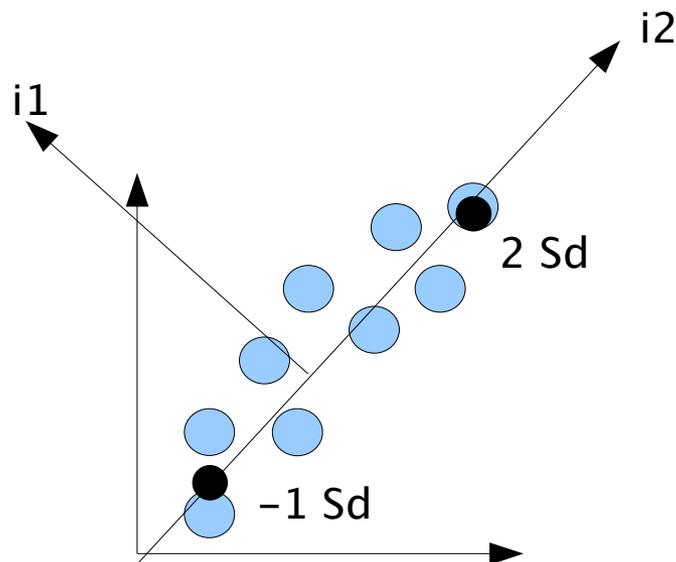
Result: Clusters

- Two separate clusters found
- Cluster “Glaucoma”:
 - 107 glaucomatous, 3 normal
- Cluster “Normal”:
 - 186 normal, 49 glaucomatous
 - Represented by only one axis
- Discussion:
 - Cluster structure correlate well with structure obtained by supervised technique

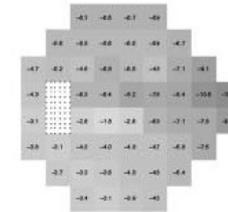


Cluster: "Glaucoma"

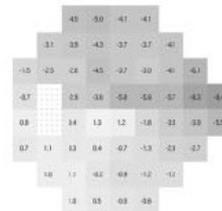
- 6 axes (i) were selected
 - criteria: reconstruction error
- Pattern Shift Along each Axis



"glaucoma" mean

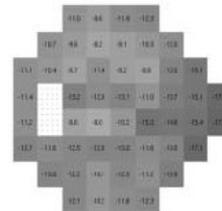


-1 SD

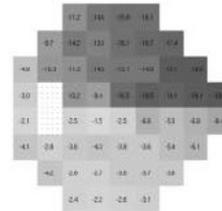
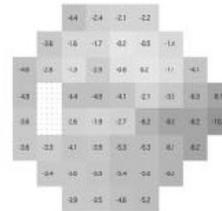


Axis 1

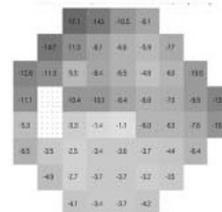
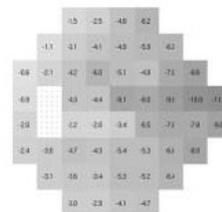
+2 SD



Axis 2



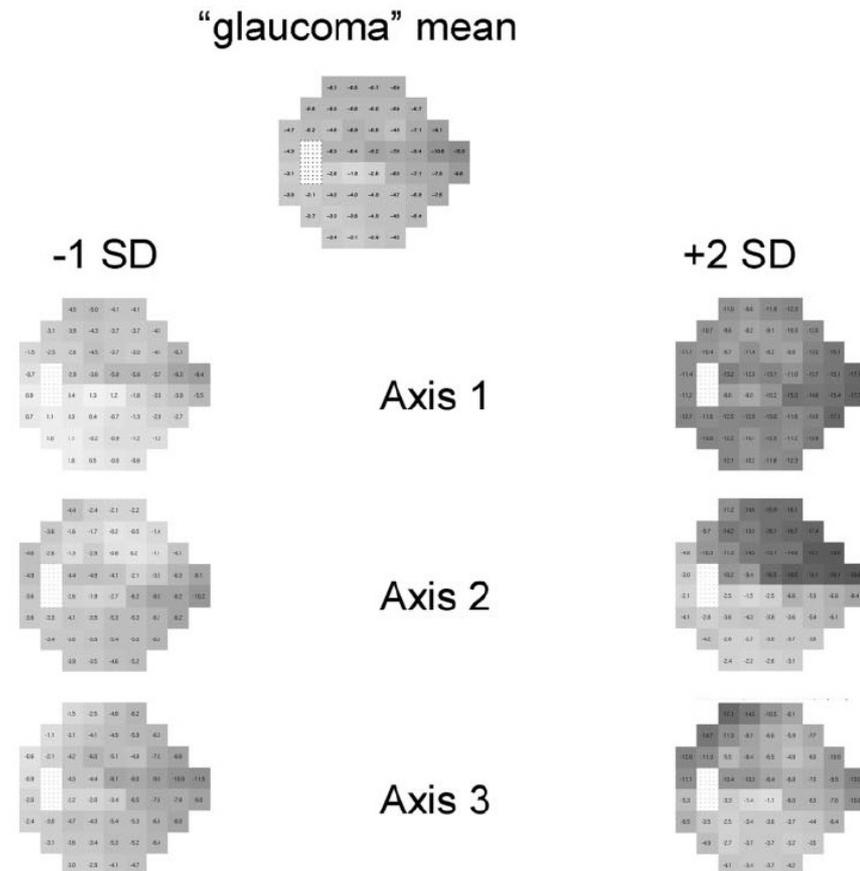
Axis 3





Cluster: “Glaucoma”

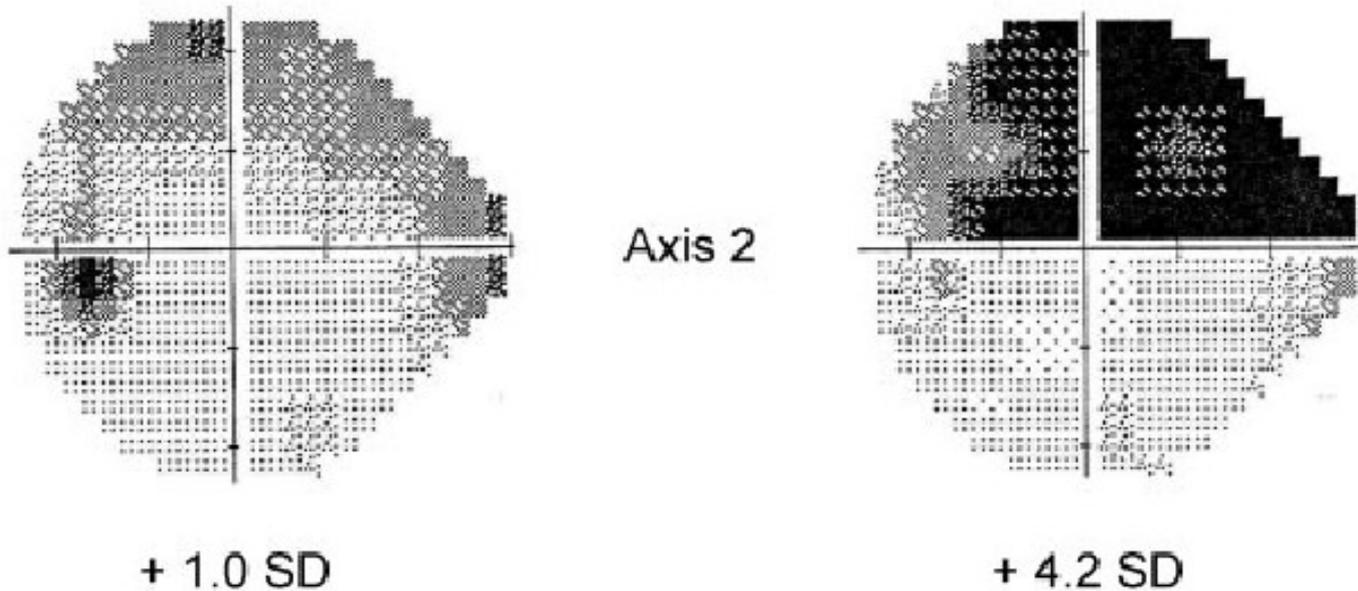
- Generated patterns represented by axis differ from each other
- Patterns resemble classic glaucoma patterns
 - Represent medical knowledge over generations
- Patterns on one axis can be considered as complementary





Cluster: "Glaucoma"

- Axis capture degree of severity





Summary and conclusion

- Unsupervised cluster and component analysis
- Extraction of meaningful
 - Clusters
 - Patterns
- ICA axes capture the sense of severity

- Conclusion:
 - (Un)supervised learning has the potential of grouping patterns that are more useful than achieved by statistical methods.
 - Unsupervised techniques might extract information still hidden to humans.
 - e.g. DTI? GRI ?