

1. Introduction

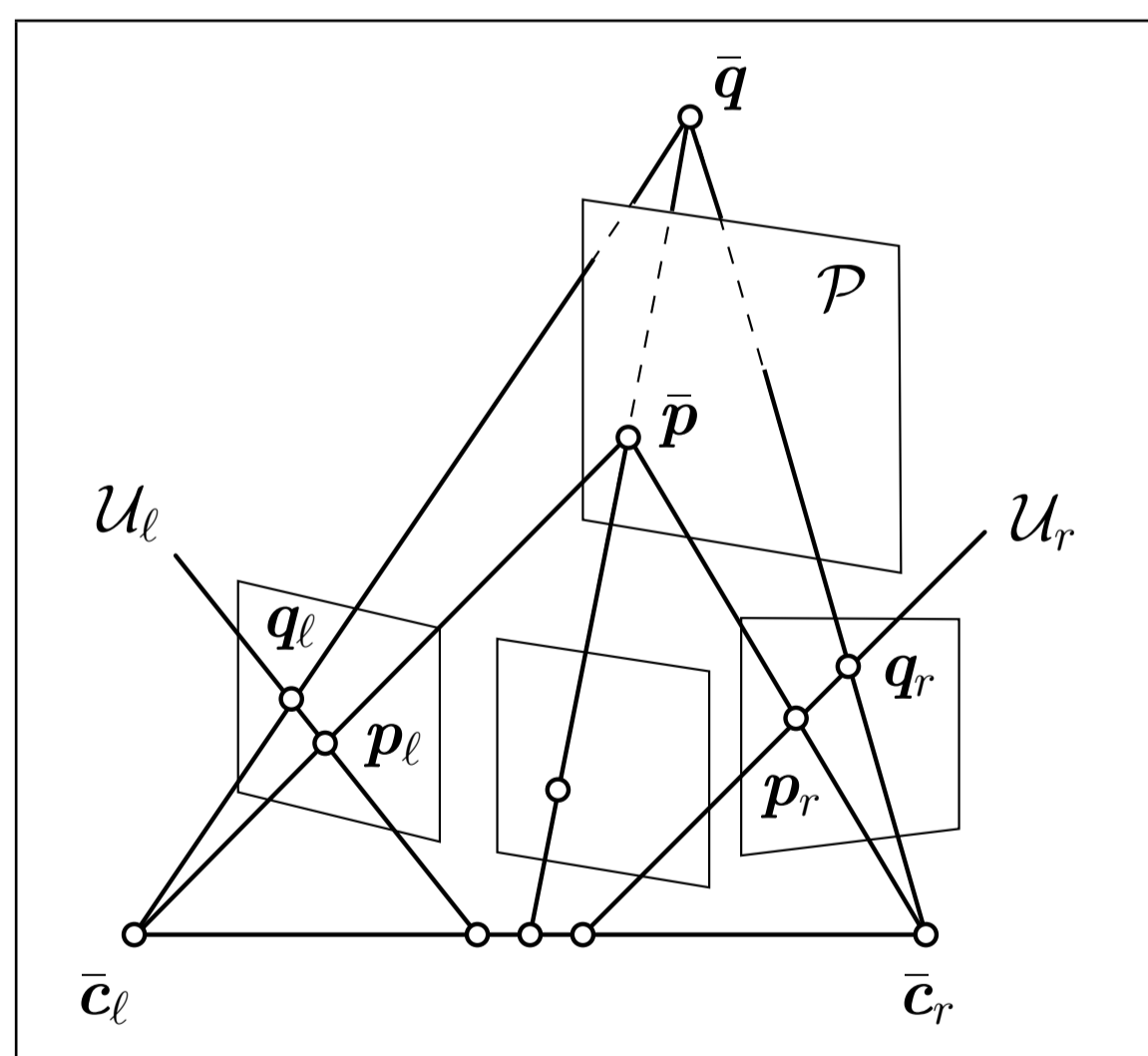
- The observed binocular disparity field is determined by the **scene structure** and the **gaze configuration**.
- Individual binocular mechanisms in area V1 are tuned to a **small range** of disparities.

2. Questions

- How does the 3-D **geometric representation** influence the observed disparity field?
- How can the **gaze component** of the disparity field be analyzed in a physiologically plausible way?

3. Representation

- The fixation point defines a **reference plane** \mathcal{P} , orthogonal to the Cyclopean gaze direction.
- The Cyclopean direction of scene point \bar{q} determines a **reference point** $\bar{p} \in \mathcal{P}$.
- The scene point \bar{q} projects to q_ℓ and q_r .
- The reference point \bar{p} projects to p_ℓ and p_r .
- The geometry is shown below, with optical centres c_ℓ and c_r , where $|c_r - c_\ell| = 1$:



- The absolute disparity, at retinal location q_ℓ , is $q_r - q_\ell$.
- The **epipolar disparity** is $d = q_\ell - p_\ell$.
- The local direction of the epipolar disparity is u_ℓ , which determines an **epipolar line** \mathcal{U}_ℓ .

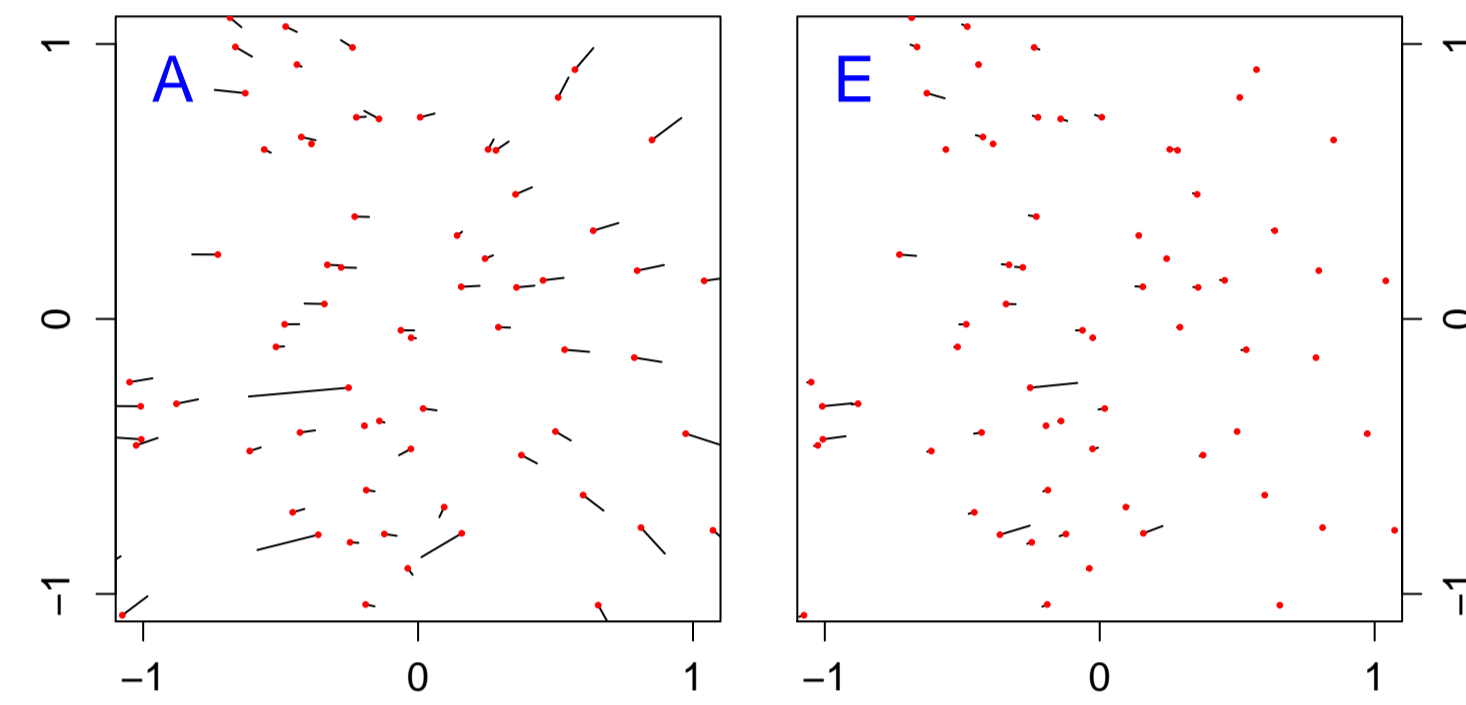
4. Experiments

- Scene Points** $i = 1 \dots M$ were Normally distributed with high variance; $M = 1000$, mean = $(0, 0, 5)^\top$, cov = $\text{diag}(100, 100, 100)$.

- Fixation points** $j = 1 \dots N$ were Normally distributed with low variance; $N = 100$, mean = $(0, 0, 5)^\top$, cov = $\text{diag}(1, 0, 1)$.

- Retinal Samples** $k = 1 \dots K$ were arranged on a Cartesian lattice; $K = 17 \times 17 = 289$.

- Two example disparity patterns (M reduced for plotting) are shown below:



- Absolute (A) and epipolar (E) disparity fields, for the same scene, plotted in the left view.

- Use of a reference plane (E) reduces the average magnitude of the disparity.

5. Sampling

- The retinal distribution of disparity information varies according to the gaze.

- Problem:** Direct smoothing introduces spurious minima in the flow, owing to the cancellation of oppositely directed vectors.

- Solution:** The disparity vector d_{ij} at point q_{ij} is represented by a rank-one 2×2 matrix:

$$D_{ij}^q = d_{ij} d_{ij}^\top$$

- The matrices encode **orientation** (mod π), rather than direction (mod 2π).

- The matrices are **smoothed** and **re-sampled** onto the retinal lattice s_k :

$$D_{jk}^s = \sum_i \alpha_{ij}^k D_{ij}^q$$

- The **spatial weighting** has an isotropic Gaussian profile:

$$\alpha_{ij}^k \propto \exp\left(-\frac{1}{2\sigma^2} |q_{ij} - s_k|^2\right)$$

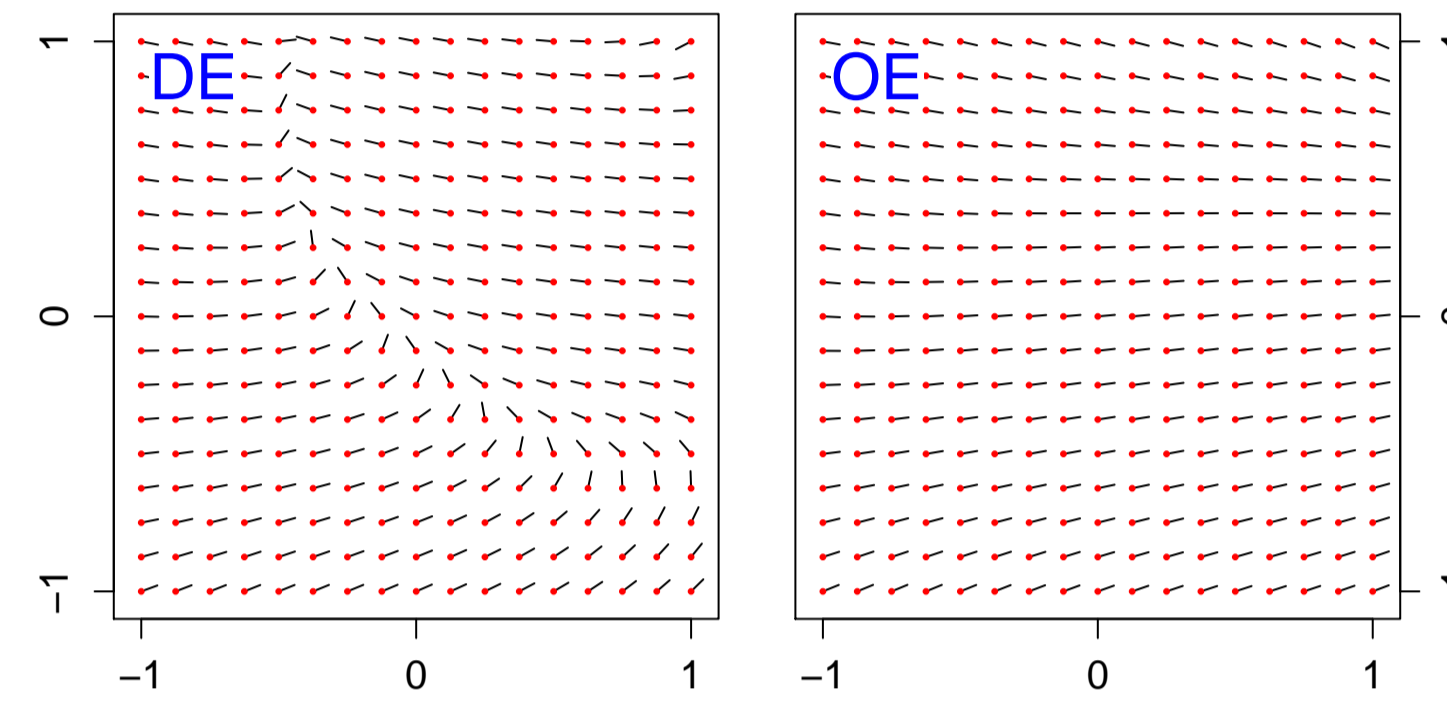
- Each average disparity matrix is rank-two, with eigenvectors u_{jk} and v_{jk} , and corresponding eigenvalues $\lambda_{jk} \geq \mu_{jk}$.

- For fixation j , the disparity matrix at retinal position s_k has the decomposition:

$$D_{jk}^s = \lambda_{jk} u_{jk} u_{jk}^\top + \mu_{jk} v_{jk} v_{jk}^\top$$

- The vectors u_{jk} approximate the epipolar geometry of fixation j , independently of the scene-structure.

- Examples of the two averaging procedures are shown here:



- Direction-average (DE), and orientation average (OE) of the epipolar disparity field shown in (E).

- Note the spurious contour in (DE), caused by averaging oppositely-directed vectors.

6. Local Structure

- Question:** What is the typical **distribution** of disparity at a given retinal point, over many fixations?

- The **gaze-averaged** flow at retinal point s_k is:

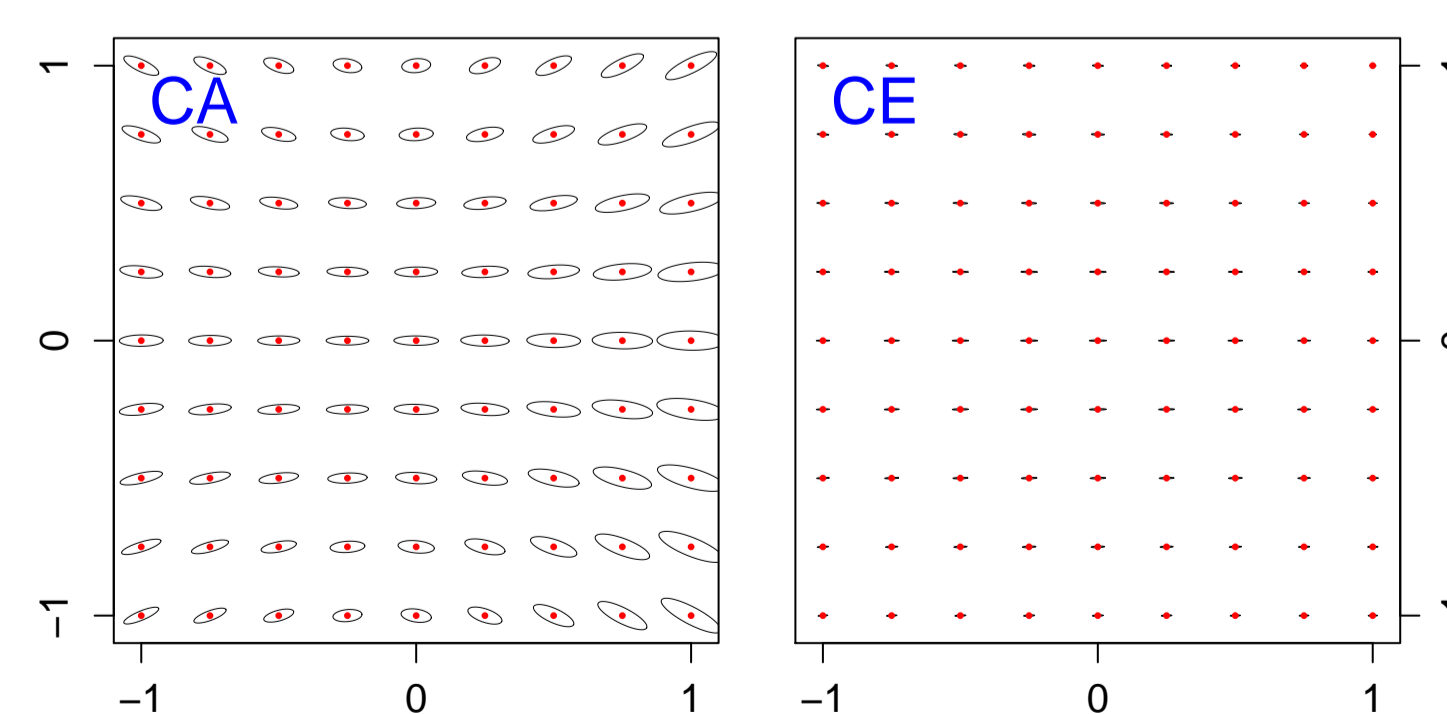
$$D_k^s = \sum_j \beta_j D_{jk}^s$$

- The gaze weighting is defined here as $\beta_j = \frac{1}{N}$.

- D_k^s encodes the disparity fluctuation at retinal position s_k , as the gaze varies.

- The inverse of D_k^s can be used to construct a **prior probability** for the local binocular correspondence.

- Iso-probability contours (K reduced for plotting) are shown below for absolute disparity (CA), and epipolar disparity (CE):



- Note that the variability is greatly reduced by the use of a reference plane (CE).

7. Global Structure

- The functional dependence between the left and right views is only approximate, after the resampling procedure.

- Hence each **binocular flow** is represented as a column-vector of length $2K + 2K$, containing u_{jk} from the left and right images:

$$f_j \in \mathbb{R}^{4K}$$

- The mean, f_0 , is subtracted from each vector.

- The N vectors are collected into a $4K \times N$ data matrix F .

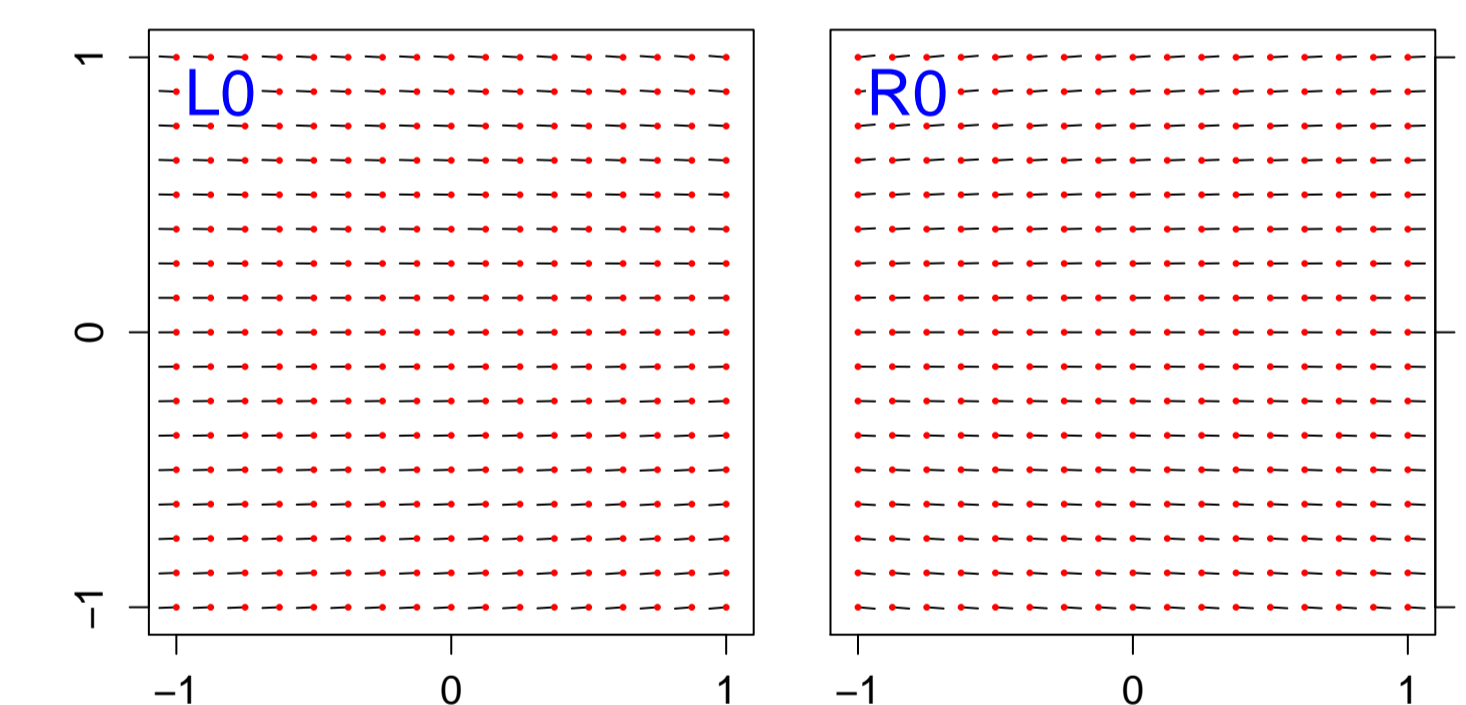
- The **singular value decomposition** of the data is

$$F = G \Sigma H^\top$$

- G is a $4K \times N$ matrix, the columns of which are **basis flows** g_j .

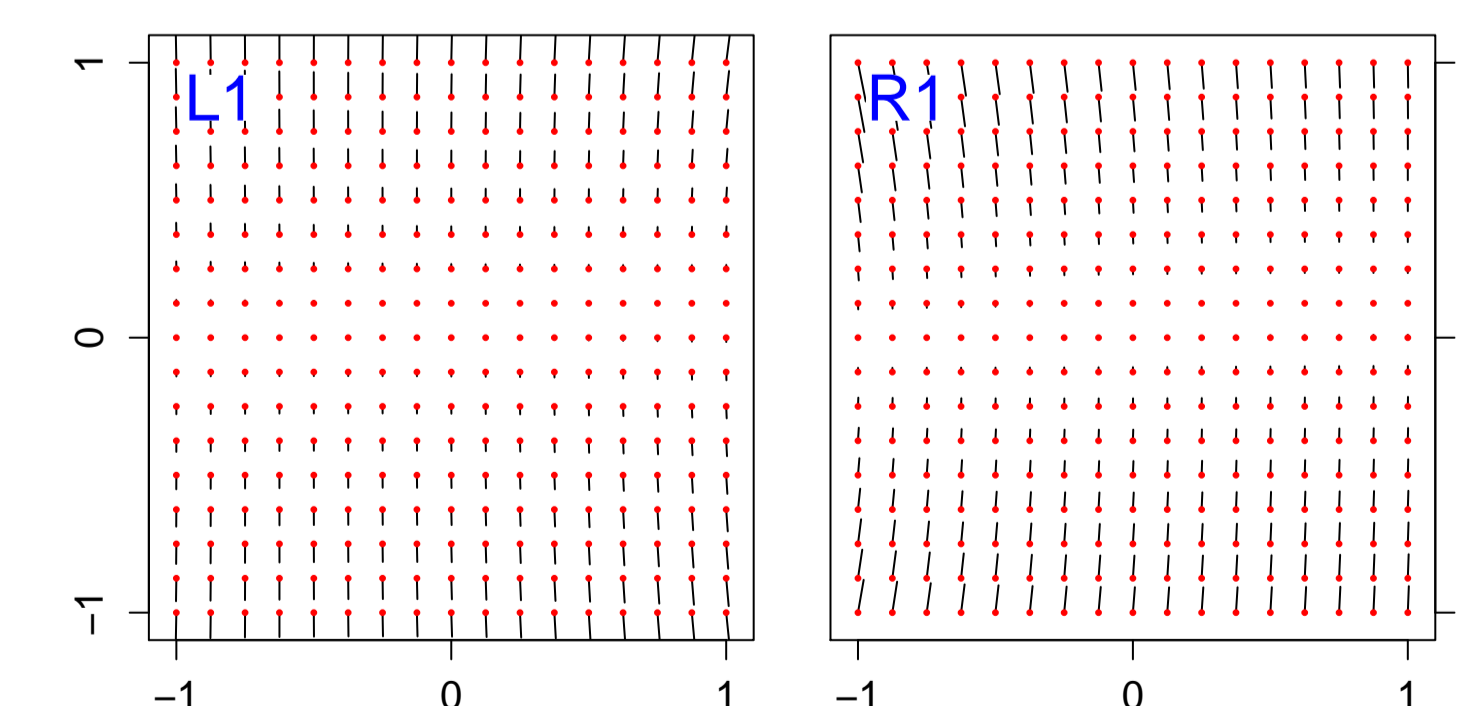
- Σ is an $N \times N$ diagonal matrix of singular values, and H^\top is an $N \times N$ orthogonal matrix.

- Left (L0) and right (R0) parts of the mean flow, f_0 , are shown below:



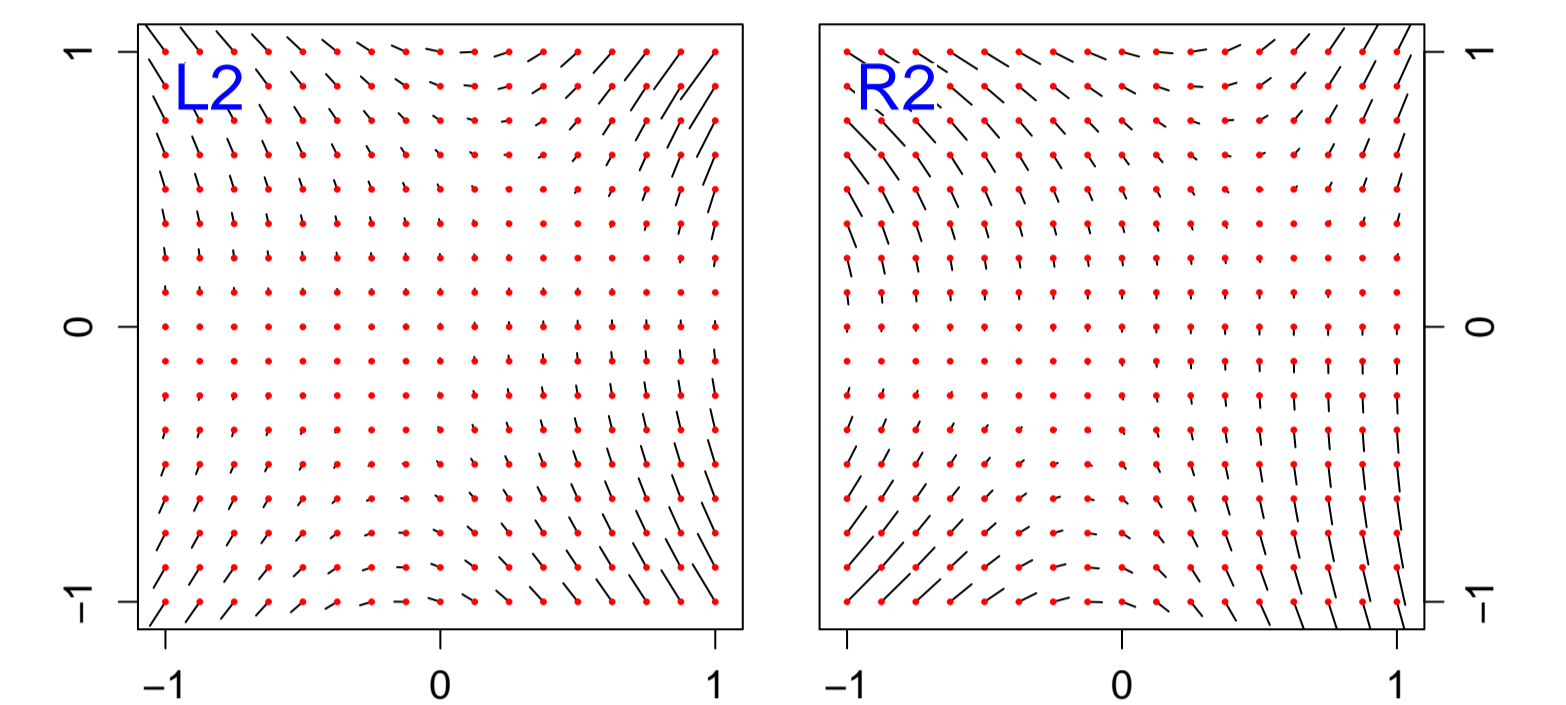
- The pair (L0,R0) represents the average epipolar geometry, over $N = 100$ random fixations.

- Left (L1) and right (R1) parts of the first basis function, g_1 , are shown below:



- The pair (L1,R1) represents a differential scaling of the disparity field, over $N = 100$ random fixations.

- Left (L2) and right (R2) parts of the second basis function, g_2 , are shown below:



- The pair (L2,R2) represents a hyperbolic deformation of the disparity field, over $N = 100$ random fixations.

8. Dimensionality

- A given flow f can be approximated by a linear combination of the first $n < N$ basis flows:

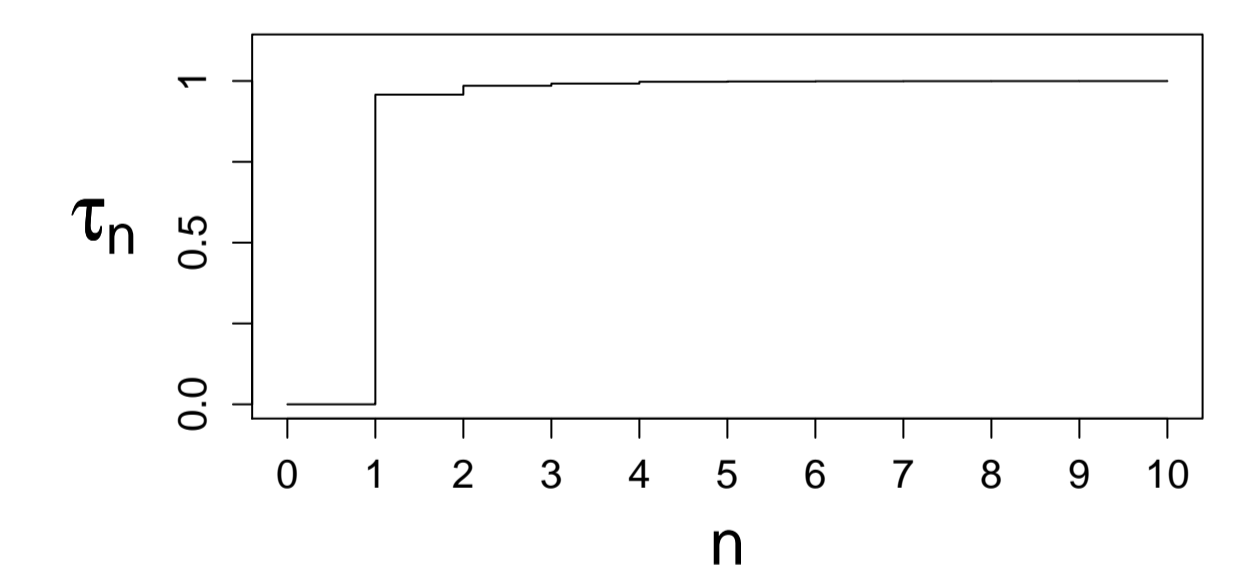
$$f \approx f_0 + \sum_j^n \gamma_j g_j$$

- The weights are projections $\gamma_j = (f - f_0)^\top g_j$.

- The proportion of **variance explained** by the first n components is:

$$\tau_n = \frac{\sum_j^n \sigma_j^2}{\sum_j^N \sigma_j^2}$$

- The proportion of gaze-variance explained by the first $n = 10$ components, with $N = 100$, is shown below:



- The **low dimensionality** of the model reflects the fixation geometry (not the scene structure).

9. Conclusions

- The **geometric representation** of disparity is important.

- The **gaze component** of retinal disparity can be locally represented.

- The distribution of the gaze component, over many binocular fixations, is low-dimensional.

10. Acknowledgments

- This work is part of the **Perception on Purpose** project, supported by EU grant 027268.