

Volume Rendering Using Principal Component Analysis

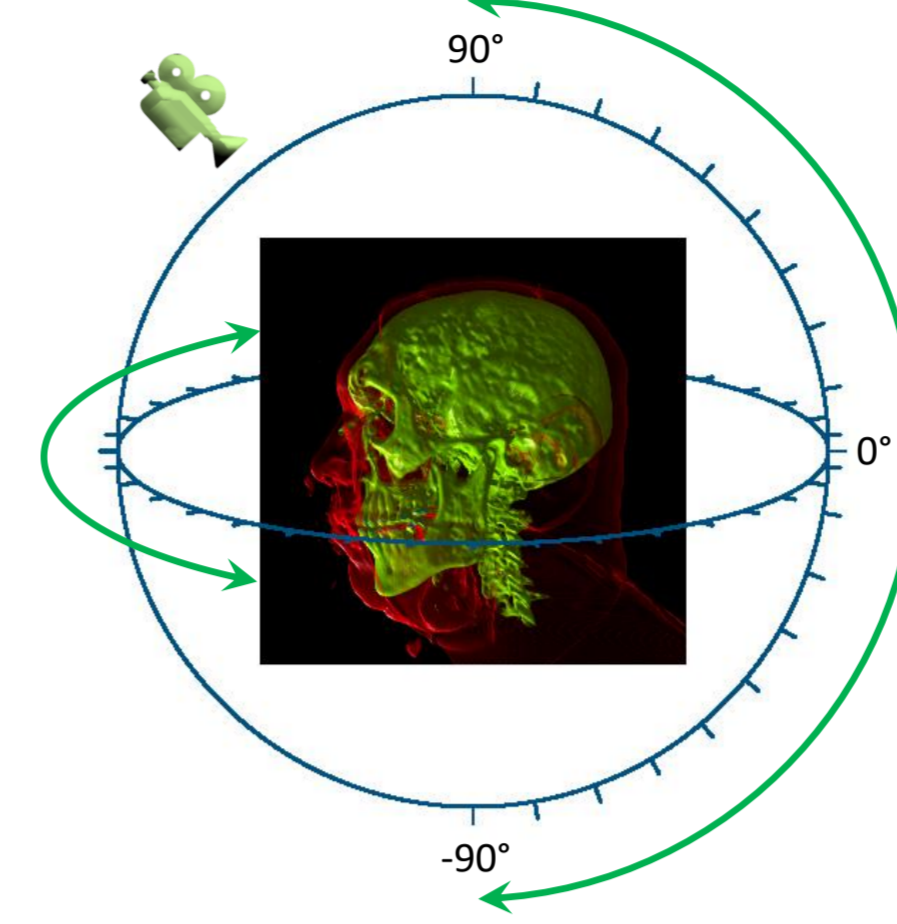
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Principal Component Analysis

- Find a set of orthogonal bases (eigenvectors) that best describe the variability in a set of input samples.
- These eigenvectors are ordered based on the variability covered by each individual eigenvector (defined by its eigenvalue).
- In our case the input samples are pre-rendered images using a standard raycaster from spherically distributed camera positions.
- Samples can be reconstructed efficiently at run-time based on their scores (projection values onto the first significant eigenvectors).
- Novel-views can be derived by interpolation of scores.



Given data samples $X = [x_1 \ x_2 \ \dots \ x_n] \in R^{d \times n}$ and eigenvector, $v \in R^d$ and, eigenvalue λ

the COVARIANCE MATRIX is defined as:

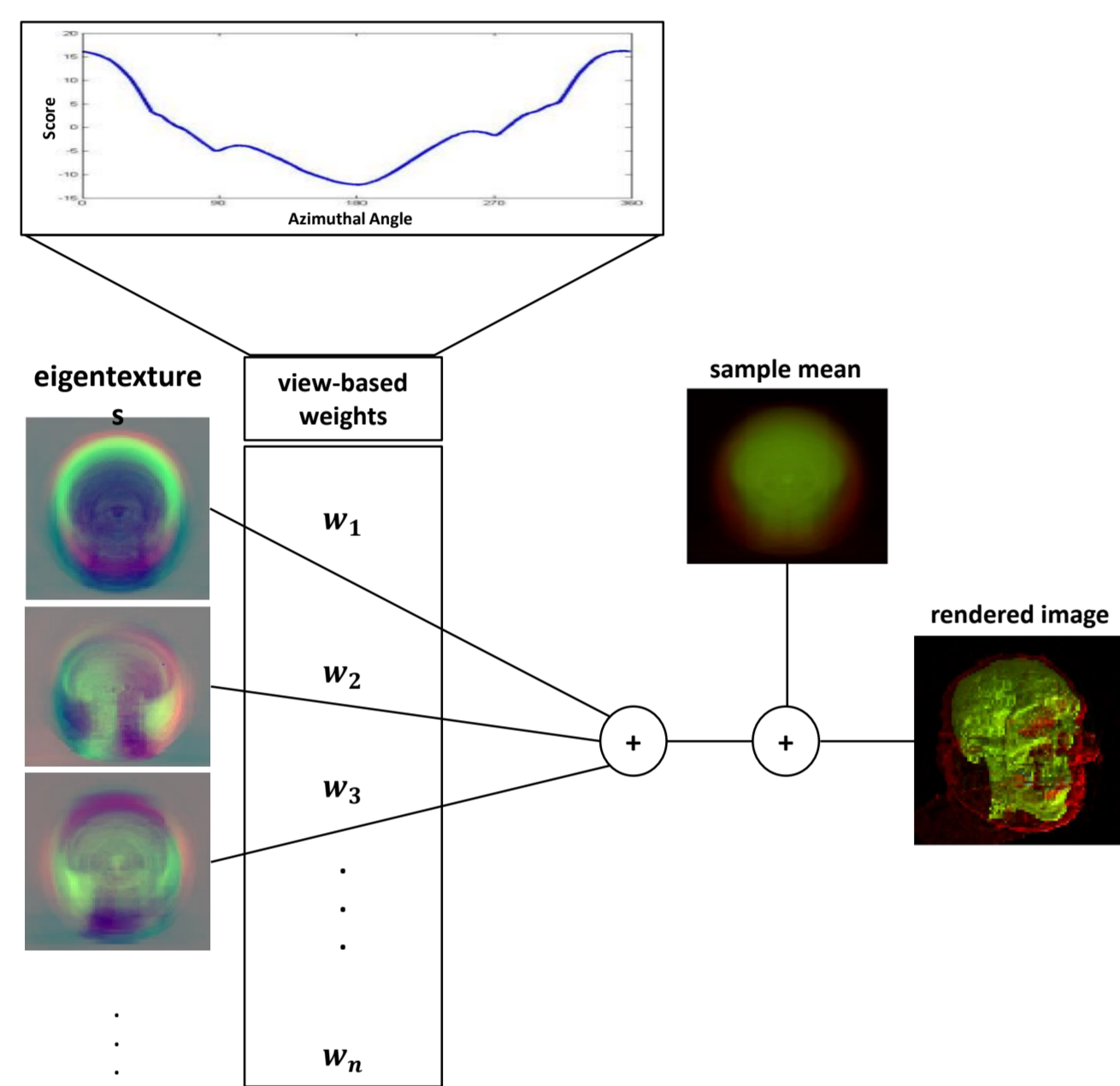
$$C = \frac{1}{n-1} X X^T = \frac{1}{n-1} \sum_{i=1}^n x_i x_i^T$$

$$(C - \lambda I)v = 0; \quad v^T v = 1$$

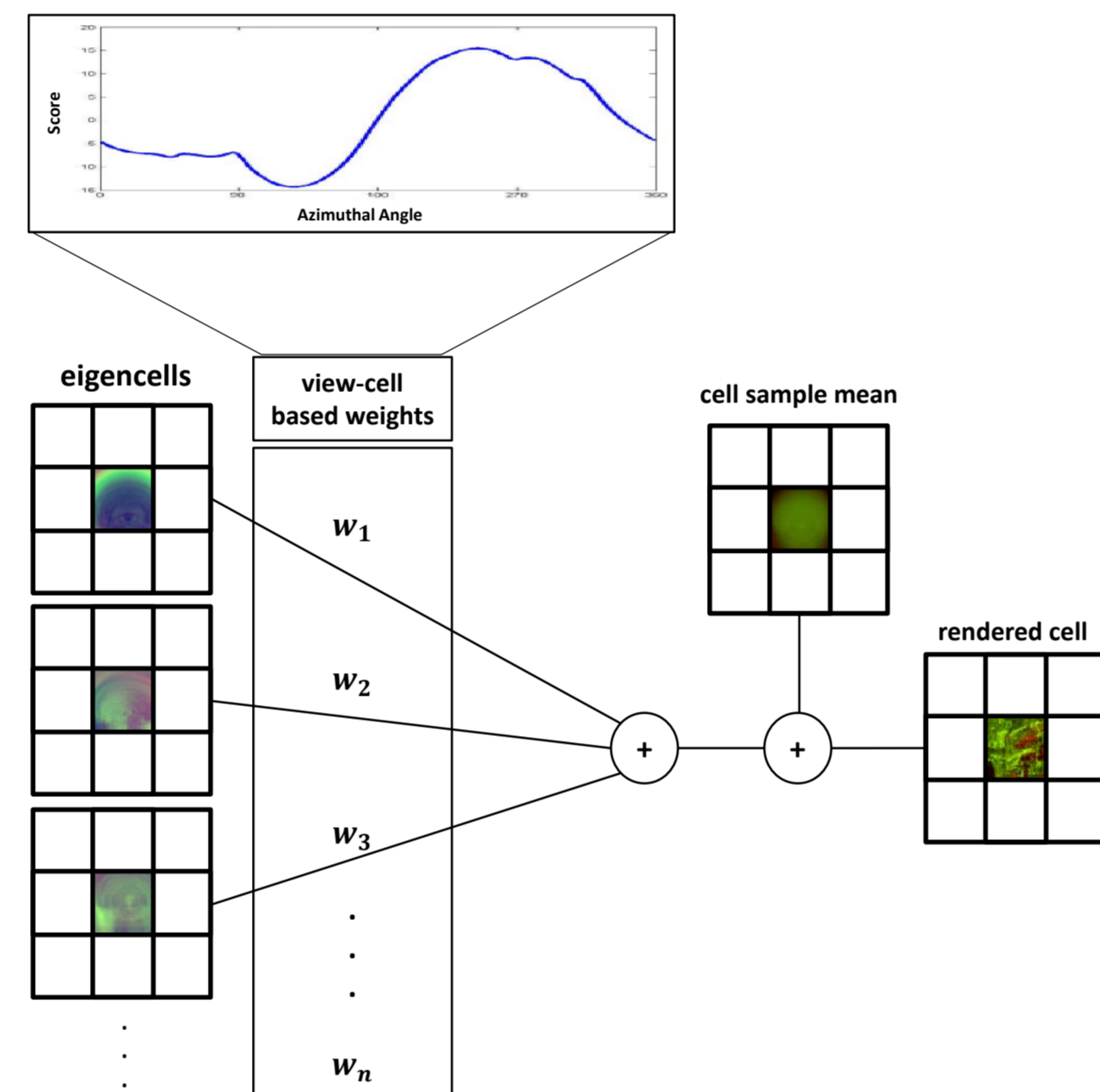
Standard PCA VS. Cell-Image PCA

PCA for Volume Rendering

Standard PCA



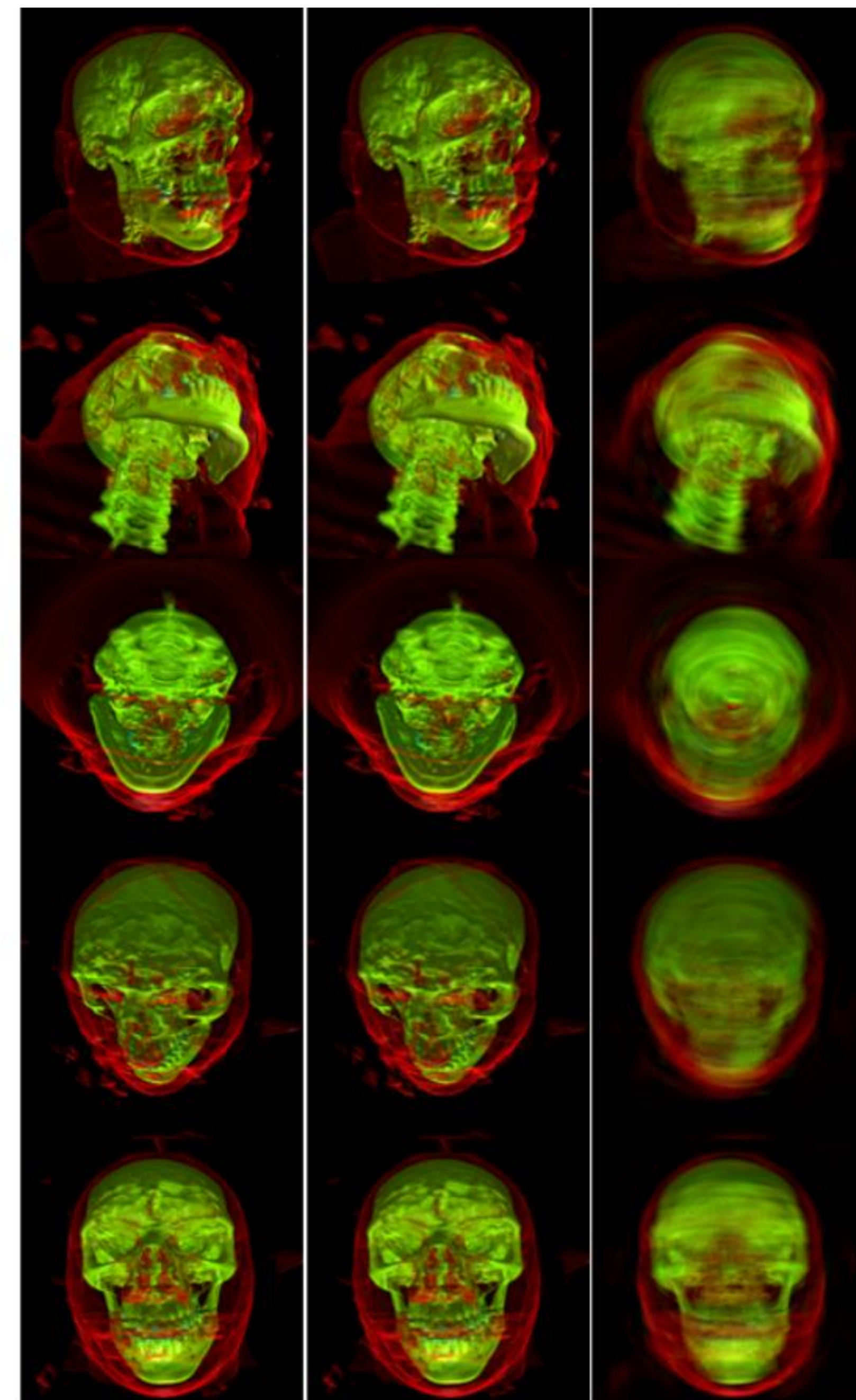
Cell-image PCA



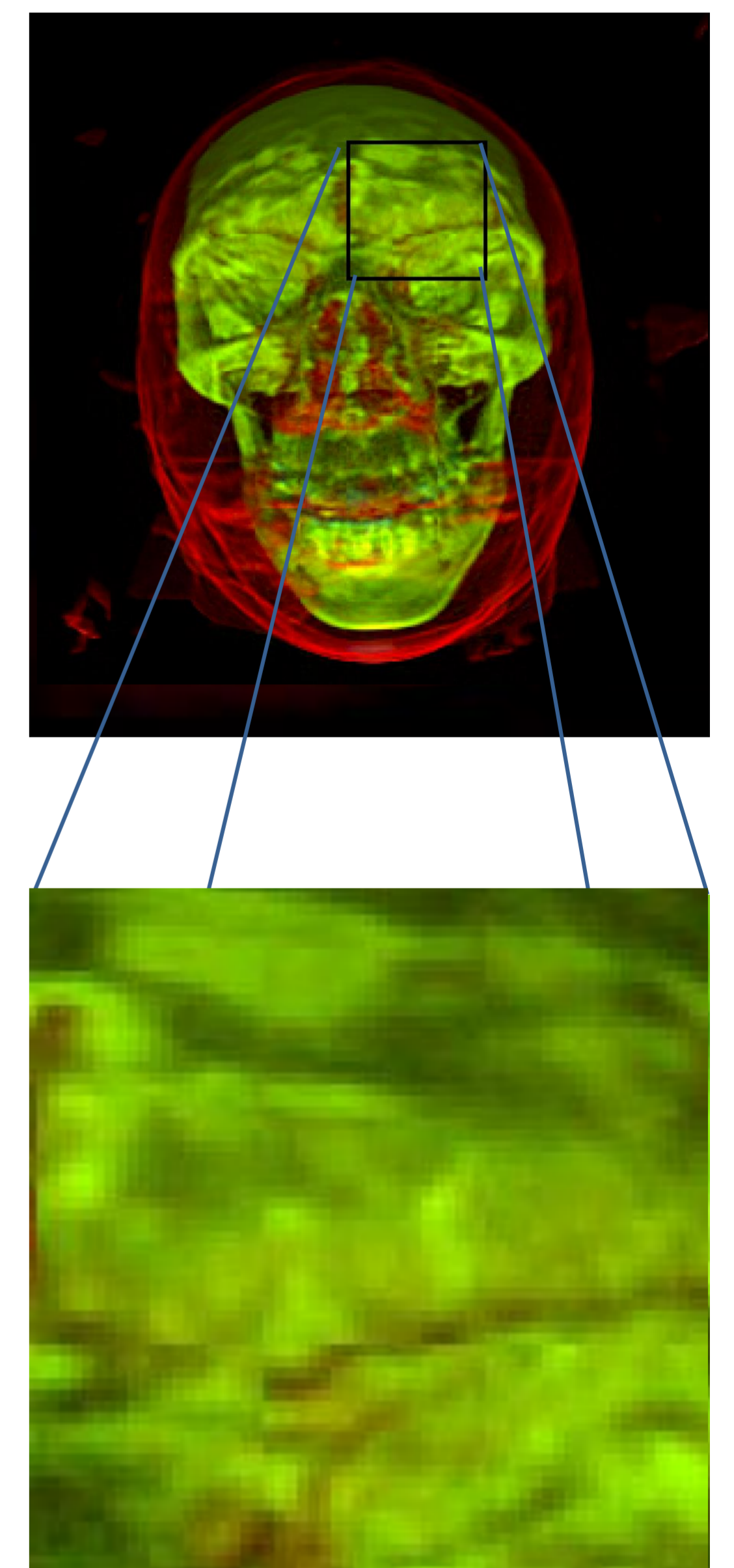
- We applied PCA to RGB images with resolution of 300x300 pixels.
- In **Cell-image PCA** the images are divided into equally sized cells and then apply PCA to each cell region individually.
- We applied cell size of 20x20 RGB pixels in **Cell-image PCA**.
- We used 100 eigenvectors per cell to encode the computed eigenspace.
- We then synthesize a novel view image by interpolating the scores in the eigenspace.

Results

Original image Cell-image PCA with 100 eigenvectors/cell Standard PCA with 100 eigenvectors



Subtle Discontinuity Artefacts with Cell-based PCA



Main Findings

Future Work

- The **Cell-image PCA** approach leads to much better quality results compared to the standard PCA.
- The **Standard PCA** results in somewhat blurry images for the same distribution of training samples.
- Cell-image PCA results in subtle discontinuity artefacts at the cell boundaries in the reconstructed images.
- A disadvantage of both PCA approaches is that changes in transfer function require recomputing the eigenspace.
- The main advantage is efficient image generation from any viewpoint. This is independent of rendering complexity, which is decoupled in the pre-processing stage.

- Conduct perceptual studies to measure the conspicuity of artefacts under different viewing and training configurations.
- Investigate strategies to ameliorate the subtle artefacts appearing in the cell boundaries for Cellimage -PCA.
- Investigate potential benefits to client-server schemes for interactive volume visualization

Acknowledgements

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