Project Develop

Workflow: Images of different times h RGB RGB Divide into grids m i m iOne gaussian distribution for each grid i $P_{t_i}(i) \sim \mathcal{N}(\mu_{ai}, \mathcal{E}_{ai}) \qquad Q_{t_i}(i) \sim \mathcal{N}(\mu_{bi}, \mathcal{E}_{bi})$ Compute distance for each grid i

 $\mathcal{O}_{\mathsf{KL}}(i) = \mathcal{A}\left(\mathcal{P}_{\mathsf{t}_1}(i), \mathcal{Q}_{\mathsf{t}_2}(i)\right) = \underbrace{\mathsf{KL}}_{\mathsf{divergence}}\left(\mathcal{P}_{\mathsf{t}_1}(i), \mathcal{Q}_{\mathsf{t}_2}(i)\right)$ obtain a map of $\mathsf{KL}_{\mathsf{divergence}}$ distance.

PKL map

Model DKI map as GMM and apply clustering for change map

Determine GMM parameters using EM algorithm Challenge: determine # of components K, grid size

Clustering by GMM:

assume $\{x_i = (Y, C, R, G, B)\}_{i=1}^n$ generated by GMM $p(x_i | \Theta) = \sum_{j=1}^{k} \langle x_j | N(x_i | \mu_j, \Xi_j) \rangle$

Z-step: $\mathcal{P}_{ij} = \frac{\mathcal{N}(i j)}{\sum_{l=1}^{K} \mathcal{N}(i, l)}$

M-step: $\chi_{\hat{j}} = \frac{1}{n} \sum_{i=1}^{n} P_{i\hat{j}}$ $\chi_{\hat{j}} = \frac{1}{n} \sum_{i=1}^{n} P_{i\hat{j}}$ Converge and choose max $\chi_{\hat{j}}$ as cluster \hat{j} .

GMM Ni = KL dvvergence value KL may histogram K components N(xi | Mj, 6j) GMM Initialization with KMeans. $\begin{cases}
\chi_{i} \}_{i=1}^{N} & \xrightarrow{k Means} \begin{cases}
\xi \chi_{i}, \chi_{z} - \chi_{i} - \chi_{N} \\
\xi C_{i}, C_{z} - C_{i} - C_{N} \end{cases}$ $\begin{cases}
M_{k} = \frac{\sum_{i=1}^{N_{k}} \chi_{i} \delta(c_{i} = k)}{N_{k}} \\
\delta_{k} = \frac{\sum_{i=1}^{N_{k}} (\chi_{i} - \mu_{k})^{2} \delta(C_{i} = k)}{N_{k}}
\end{cases}$ $\chi_{k} = \frac{N_{k}}{N}$

GMM Convergence Metric using log likelihood

$$\ln P(X|\mu, 6, d) = \sum_{n=1}^{N} \ln \sum_{k=1}^{K} d_k N(X_n|\mu_k, 6_k)$$