Towards sparse coding of rainfall spectral atoms: A TRMM legacy

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Outline

ShARP: latest update

- A demo MATLAB code will be released next week.

- A new paper: On evaluation of ShARP over coastal zones and snow-covered lands.

Coding rainfall spectral atoms

- Guiding physically-based retrievals
- New directions in retrieval methods

Shrunken locally linear embedding Algorithm for Retrieval of Precipitation



Dictionaries:

Tb: $\mathbf{B} = [\mathbf{b}_1 | \mathbf{b}_2 | \dots | \mathbf{b}_{M-1} | \mathbf{b}_M] \in \mathfrak{R}^{n_c \times M}$

Rainfall: $\mathbf{R} = [\mathbf{r}_1 | \mathbf{r}_2 | \dots | \mathbf{r}_{M-1} | \mathbf{r}_M] \in \mathfrak{R}^{n_r \times M}$

where $\mathbf{r}_{i} = (r_{1i}, r_{2i}, ..., r_{n_{r}i})^{T}$ and $\mathbf{b}_{i} = (b_{1i}, b_{2i}, ..., b_{n_{c}i})^{T}$

ShARP: Some Details

Detection Step:

- **o** for a given $\mathbf{y} \in \mathfrak{R}^{n_r}$ and the dictionary pairs $\mathbf{B} = {\{\mathbf{b}_i\}}_i^M$, $\mathbf{R} = {\{\mathbf{r}_i\}}_i^M$
- find TB sub-dictionary: $\mathbf{B}_{\mathcal{S}} = [\mathbf{b}_1 | \mathbf{b}_2 | \dots | \mathbf{b}_K] = \{\mathbf{b}_k\}_{k=1}^K$, where \mathbf{b}_k are the *k*-nearest neighbors of \mathbf{y}
- o obtain the rainfall sub-dictionary: $\mathbf{R}_{\mathcal{S}} = [\mathbf{r}_1 | \mathbf{r}_2 | \dots | \mathbf{r}_{\mathcal{K}}] = {\{\mathbf{r}_k\}}_{k=1}^{\mathcal{K}}$
- o rain/no-raining detection by a probabilistic vote rule

• Estimation Step:

$$\begin{array}{ll} \underset{\mathbf{c}}{\text{minimize}} & \left\| \mathbf{W}^{1/2} \left(\mathbf{y} - \mathbf{B}_{\mathcal{S}} \mathbf{c} \right) \right\|_{2}^{2} + \lambda_{1} \left\| \mathbf{c} \right\|_{1} + \lambda_{2} \left\| \mathbf{c} \right\|_{2}^{2} \\ \text{subject to} & \mathbf{c} \succeq \mathbf{0}, \ \mathbf{1}^{\mathrm{T}} \mathbf{c} = \mathbf{1}. \end{array}$$

o
$$\ell_2$$
-norm: $\|\mathbf{c}\|_2^2 = \Sigma_i |c_i|^2$ and ℓ_1 -norm: $\|\mathbf{c}\|_1 = \Sigma_i |c_i|$.
o estimate of rainfall profile: $\hat{\mathbf{x}} = \mathbf{R}_S \hat{\mathbf{c}}$.

ShARP: storm-scale

Cyclone Sidr (10/15/2007), 13:59 UTC



► Study area, annual MODIS snow cover and TMI-Scattering Index in 2013



- ▶ Probability of hit (all overpasses in 2013 while PR is the reference over land)
- Probability of false alarm



▶ Oct-Nov-Dec (2013)



► Jul-Aug-Sep (2013)



▶ Oct-Nov-Dec (2013)



► Jul-Aug-Sep (2013)



Error metrics (2013)



1- Ebtehaj, A.M., R.L. Bras, E. Foufoula-Georgiou (2015), Shrunken locally linear embedding for passive microwave retrieval of precipitation, *IEEE Trans. on Geosci. and Remote Sens.*vol 53(7), doi:10.1109/TGRS.2014.2382436.

2- Ebtehaj, A.M., R.L. Bras, E. Foufoula-Georgiou (2015), On evaluation of ShARP passive rainfall retrievals over snow-covered land surfaces and coastal zones, arxiv.org/abs/1503.05495.

Why Rainfall Atoms?

► Goals

- o guiding physically based retrieval approaches.
- o new directions for precipitation retrieval.
- ► Idea:





How can we learn a compact set of rainfall spectral atoms?

Spectral Atoms and a Rain/No-rain Detection Algorithm

notations:

pixel-level spectral observations: $\mathbf{y} = [\mathsf{Tb}_{10v}, \dots, \mathsf{Tb}_{85h}]^T \in \mathfrak{R}^9$ rainfall profile: $\mathbf{x} = [r_1, r_2, \dots, r_n]^T \in \mathfrak{R}^n$ raining (r) and dry (d) spectral atoms: $\mathbf{D}_{\mathrm{T}}^{\{r,d\}} = [\mathbf{d}_{\mathrm{T1}}|\mathbf{d}_{\mathrm{T2}}|\dots|\mathbf{d}_{\mathrm{TK}}] \in \mathfrak{R}^{9 \times K}$

estimating raining and dry atoms via sparse coding

$$\begin{array}{ll} \underset{\mathbf{D}_{\mathrm{T}},\mathbf{c}_{i}}{\text{minimize}} & \sum_{i=1}^{M} \left\| \mathbf{y}_{i}^{\{\mathrm{r},\,\mathrm{d}\}} - \mathbf{D}_{\mathrm{T}}^{\{\mathrm{r},\,\mathrm{d}\}} \mathbf{c}_{i} \right\|_{2}^{2} \\ \text{subject to} & \left\| \mathbf{c}_{i} \right\|_{0} \leq k_{0}, \ \mathbf{c}_{i} \succeq 0, \ \mathbf{1}^{\mathrm{T}} \mathbf{c}_{i} = 1 \end{array}$$

rain/no-rain detection:

$$\begin{split} j_{\mathbf{y}} &= \underset{j \in \{r, d\}, \mathbf{c}}{\operatorname{argmin}} \quad \left\| \mathbf{y} - \mathbf{D}_{\mathrm{T}}^{j} \mathbf{c} \right\|_{2}^{2} \\ & \text{subject to} \quad \mathbf{c} \succeq 0, \ \mathbf{1}^{\mathrm{T}} \mathbf{c} = 1. \end{split}$$
where the set {r, d} denotes the raining (r) and dry (d) classes.

Channels and Number of Atoms

► Over land (22-85 GHz) — Over ocean (1-85 GHz)



• Number of the rainfall atoms: $\sim 2 \times$ number of channels

- o K=10 rainfall atoms over land
- o K=18 rainfall atoms over ocean

Rainfall Spectral Atoms – Over Ocean

 Obtained from 250 randomly sampled orbits in 5 years of TMI-PR overlapping obs.



Rainfall Spectral Atoms – Over Land

 Obtained from 250 randomly sampled orbits in 5 years of TMI-PR overlapping obs.



A Detection Experiment–Ocean

► A storm track - 1998/06/28.03357







New Alg.

A Detection Experiment–Land

► A storm track - 1998/06/28.03357













Earth Surface Radiation Classes using TMI Overpasses

Rainfall atoms for different earth surface types (currently working)

BT segmentation (Euclidean)



Land Surface Emissivity Classes



Emissivity Segmentation

AMSR-E emissivity data by Hamid Norouzi at CUNY.

Current and Future Works

- ► Obtaining the spectral atoms for different land surface classes.
- Providing physical insights by studying the precipitation vertical profile of the spectral atoms.
- Designing the estimation part, emphasizing on low and extreme rainfall.
- ► Studying rainfall eigen-spectrum via PCA.

Thank You