# Combining Scattering Transform paths

# cuts off the exponential growth while

retaining key features

### Collating Transform David Weber, Naoki Saito UC Davis

#### **Scattering Transform**

A Scattering layer consists of a collection of frames  $\Psi^{\ell} = \{\phi\} \bigcup \{\psi_{\lambda^{\ell}}\}_{\lambda^{\ell} \in \Lambda^{\ell}}, \text{ a nonlinearity } \sigma \text{ such as}$ ReLU, and a subsampler  $R_r$  with rate r:  $U[\lambda^{\ell}](f) = R_r \sigma(\psi_{\lambda^{\ell}} \star f)$ 

### Collating Transform



#### Comparison

Learned CNNs match the structure of the collating transform, but learn the entire frame collator  $A_{ki}\psi_{\lambda^{\ell}}$  product tensor, rather than just the collation.

Given *F* frames in each layer, *c* input channels, and  $n^d$  coefficients in each channel, there are  $cF^L$  scattering paths at layer *L*, and  $\sum_{\ell=1}^{L} cn^d \left(\frac{F}{r^d}\right)^{\ell}$ output coefficients. If there are  $K_\ell$  collators at layer  $\ell$ , there are  $FK_\ell$  nodes, and a total of

A scattering path  $p = (\lambda^1, ..., \lambda^L)$  consists of repeatedly applying these operations and finally averaging:

 $U^{\ell}[p](f) = U[\lambda^{\ell}] \cdots U[\lambda^{1}](f)$  $S[p](f) = \phi \star U[p](f)$ 

#### **Collating Transform**

Given a set of signals  $f_1, \ldots, f_c$ , a collating layer interposes a sum across frames using a collating matrix  $A \in \mathbb{R}^{K \times c}$ :

$$C[\lambda^{\ell}, k](f_1, \dots, f_c) = R_r \sigma \left( \sum_{i=1}^c A_{ki} \psi_{\lambda^{\ell}} \star f_i \right)$$
$$S[\lambda^{\ell}, k](f) = \phi \star C[\lambda^{\ell}, k](f)$$

#### Theoretical properties of both

The following carry over simply from the scattering transform to the collating transform:

#### $\sum_{\ell=0}^{-} \frac{n^d}{r^{d\ell}} F K_{\ell} \text{ coefficients.}$

#### **Combining Paths**

If path *a* contributes values at  $x_a$ , and path *b* at  $x_b$ , then if  $\{x_a \pm (2\Phi + A)\}$  and  $\{x_B \pm (2\Phi + B)\}$  don't overlap, adding them keeps the same value at those locations, using the sparsity inducing properties of frames.



**Theorem 1.** *Translation decays exponentially with depth:* 

 $S[\lambda^{\ell}, k](T_{c}f) = T_{c/r_{1}...r_{\ell}}S[\lambda^{\ell}, k](f)$ 

**Theorem 2.** Given upper frame bound  $B_{\ell}$ , lipschitz bound  $\gamma_{\ell}$  on  $\sigma$ , and c signals of dimension d, if we have  $\max\{B_{\ell}, B_{\ell}r_{\ell+1}^{-d}\gamma_{\ell+1}(\max_{ji}|A_{ji}^{\ell}|^2)\} \leq 1$ , then there is a universal C so for R band-limited functions  $f \in \mathscr{L}^2(\mathbb{R}^d, \mathbb{R}^c), \omega \in \mathscr{C}(\mathbb{R}^d, \mathbb{R}^c), \tau \in \mathscr{C}^1(\mathbb{R}^d, \mathbb{R}^d),$ 

 $\left(F_{\tau,\omega}f\right)_{i} \coloneqq e^{i\omega_{i}(x)}f_{i}(x+\tau(x))$  $\left\|S\left[F_{\tau,\omega}f\right] - S[f]\right\| \leq C\left(R\|\tau\|_{\infty} + \max_{j}\|\omega_{j}\|_{\infty}\right)\|f\|_{2}$ 



Original

For 2D transforms, there is also a spatial element to overlap



Extends to groups of paths that use the same frame at a given layer, decomposition can happen at different levels.