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## 21 Chapter 2

# 22 Convolutions on graph domains

## 23 Introduction

24 Defining a convolution of signals over graph domains is a challenging problem.  
25 Obviously, if the graph is not a grid graph there exists no natural definition.  
26 We first analyse the reasons why the euclidean convolution operator is useful  
27 in deep learning, and give a characterization. Then we will search for domains  
28 onto which a convolution with these properties can be naturally obtained.  
29 This will lead us to put our interest on representation theory and convolutions  
30 defined on groups. As the euclidean convolution is just a particular case of  
31 the group convolution, it makes perfect sense to steer our construction in  
32 this direction. Hence, we will aim at transferring its representation on the  
33 vertex domain. First we will do this construction agnostically of the edge  
34 set. Then, we will introduce the role of the edge set and see how it should  
35 influence it. This will provide us with some particular classes of graphs for  
36 which we will obtain a natural construction with the wanted characteristics  
37 that we exposed in the first place. Finally, we can relax some aspect of the  
38 construction to adapt it to graphs that are not order-regular. The obtained  
39 construction is a set of general expressions for convolutions on graph domains,  
40 which preserve some key properties.

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## 2.1 Analysis of the classical convolution

In this section, we are exposing a few properties of the classical convolution that a generalization to graphs would likely try to preserve. For now let's consider a graph  $G$  agnostically of its edges *i.e.*  $G \cong V$  is just the set of its vertices.

### 2.1.1 Properties of the convolution

Consider an edge-less grid graph *i.e.*  $G \cong \mathbb{Z}^2$ . By restriction to compactly supported signals, this case encompass the case of images.

#### Definition 1. Convolution on $\mathcal{S}(\mathbb{Z}^2)$

Recall that the (discrete) convolution between two signals  $s_1$  and  $s_2$  over  $\mathbb{Z}^2$  is a binary operation in  $\mathcal{S}(\mathbb{Z}^2)$  defined as:

$$\forall (a, b) \in \mathbb{Z}^2, (s_1 * s_2)[a, b] = \sum_i \sum_j s_1[i, j] s_2[a - i, b - j]$$

#### Definition 2. Convolution operator

A *convolution operator* is a function of the form  $f_w : x \mapsto x * w$ , where  $x$  and  $w$  are signals of domains for which the convolution  $*$  is defined. When  $*$  is not commutative, we differentiate the *right-action* operator  $x \mapsto x * w$  from the *left-action* one  $x \mapsto w * x$ .

The following properties of the convolution on  $\mathbb{Z}^2$  are of particular interest for our study.

#### Linearity

Operators produced by the convolution are linear. So they can be used as linear parts of layers of neural networks.

#### 84 **Locality and weight sharing**

85 When  $w$  is compactly supported on  $K$ , an impulse response  $f_w(x)[a, b]$  amounts  
 86 to a  $w$ -weighted aggregation of entries of  $x$  in a neighbourhood of  $(a, b)$ , called  
 87 the *local receptive field*.

#### 88 **Commutativity**

89 The convolution is commutative. However, it won't necessarily be the case  
 90 on other domains.

#### 91 **Equivariance to translations**

92 Convolution operators are equivariant to translations. Below, we show that  
 93 the converse of this result also holds with Proposition 6.

### 94 **2.1.2 Characterization on grid graphs**

95 Let's recall first what is a transformation, and how it acts on signals.

#### 96 **Definition 3. Transformation**

97 A *transformation*  $f : V \rightarrow V$  is a function with same domain and codomain.  
 98 The set of transformations is denoted  $\Phi(V)$ . The set of bijective transforma-  
 99 tions is denoted  $\Phi^*(V) \subset \Phi(V)$ .

100 In particular,  $\Phi^*(V)$  forms the symmetric group of  $V$  and can move signals  
 101 of  $\mathcal{S}(V)$  by linear extension of its group action.

#### 102 **Lemma 4. Extension to $\mathcal{S}(V)$ by group action**

103 An transformation  $f \in \Phi^*(V)$  can be extended linearly to the signal space  
 104  $\mathcal{S}(V)$ , and we have:

$$\forall s \in \mathcal{S}(V), \forall v \in V, f(s)[v] := L_f(s)[v] = s[f^{-1}(v)]$$

105 *Proof.* Let  $s \in \mathcal{S}(V)$ ,  $f \in \Phi^*(V)$ ,  $L_f \in \mathcal{L}(\mathcal{S}(V))$  s.t.  $\forall v \in V, L_f(\delta_v) = \delta_{f(v)}$ .

106 Then, we have:

$$\begin{aligned} L_f(s) &= \sum_{v \in V} s[v] L_f(\delta_v) \\ &= \sum_{v \in V} s[v] \delta_{f(v)} \end{aligned}$$

$$\text{So, } \forall v \in V, L_f(s)[v] = s[f^{-1}(v)]$$

107

□

108 We also recall the formalism of translations.

109 **Definition 5. Translation on  $\mathcal{S}(\mathbb{Z}^2)$**

110 A translation on  $\mathbb{Z}^2$  is defined as a transformation  $t \in \Phi^*(\mathbb{Z}^2)$  such that

$$\exists(a, b) \in \mathbb{Z}^2, \forall(x, y) \in \mathbb{Z}^2, t(x, y) = (x + a, y + b)$$

111 It also acts on  $\mathcal{S}(\mathbb{Z}^2)$  with the notation  $t_{a,b}$  i.e.

$$\forall s \in \mathcal{S}(\mathbb{Z}^2), \forall(x, y) \in \mathbb{Z}^2, t_{a,b}(s)[x, y] = s[x - a, y - b]$$

112 For any set  $E$ , we denote by  $\mathcal{T}(E)$  its translations if they are defined.

113 The next proposition fully characterizes convolution operators with their  
114 translational equivariance property. This can be seen as a discretization of a  
115 classic result from the theory of distributions (Schwartz, 1957).

116 **Proposition 6. Characterization of convolution operators on  $\mathcal{S}(\mathbb{Z}^2)$**

117 On real-valued signals over  $\mathbb{Z}^2$ , the class of linear transformations that are  
118 equivariant to translations is exactly the class of convolutive operations i.e.

$$\exists w \in \mathcal{S}(\mathbb{Z}^2), f = . * w \Leftrightarrow \begin{cases} f \in \mathcal{L}(\mathcal{S}(\mathbb{Z}^2)) \\ \forall t \in \mathcal{T}(\mathcal{S}(\mathbb{Z}^2)), f \circ t = t \circ f \end{cases}$$

119

120 *Proof.* The result from left to right is a direct consequence of the definitions:

$$\begin{aligned}
& \forall s \in \mathcal{S}(\mathbb{Z}^2), \forall s' \in \mathcal{S}(\mathbb{Z}^2), \forall (\alpha, \beta) \in \mathbb{R}^2, \forall (a, b) \in \mathbb{Z}^2, \\
& f_w(\alpha s + \beta s')[a, b] = \sum_i \sum_j (\alpha s + \beta s')[i, j] w[a - i, b - j] \\
& = \alpha f_w(s)[a, b] + \beta f_w(s')[a, b] \quad (\text{linearity}) \\
& \forall s \in \mathcal{S}(\mathbb{Z}^2), \forall (\alpha, \beta) \in \mathbb{Z}^2, \forall (a, b) \in \mathbb{Z}^2, \\
& f_w \circ t_{\alpha, \beta}(s)[a, b] = \sum_i \sum_j t_{\alpha, \beta}(s)[i, j] w[a - i, b - j] \\
& = \sum_i \sum_j s[i - \alpha, j - \beta] w[a - i, b - j] \\
& = \sum_{i'} \sum_{j'} s[i', j'] w[a - i' - \alpha, b - j' - \beta] \quad (1) \\
& = f_w(s)[a - \alpha, b - \beta] \\
& = t_{\alpha, \beta} \circ f_w(s)[a, b] \quad (\text{equivariance})
\end{aligned}$$

121 Now let's prove the result from right to left.

122 Let  $f \in \mathcal{L}(\mathcal{S}(\mathbb{Z}^2))$ ,  $s \in \mathcal{S}(\mathbb{Z}^2)$ . We suppose that  $f$  commutes with trans-  
 123 lations. Recall that  $s$  can be linearly decomposed on the infinite family of  
 124 dirac signals:

$$s = \sum_i \sum_j s[i, j] \delta_{i, j}, \text{ where } \delta_{i, j}[x, y] = \begin{cases} 1 & \text{if } (x, y) = (i, j) \\ 0 & \text{otherwise} \end{cases}$$

125 By linearity of  $f$  and then equivariance to translations:

$$\begin{aligned}
f(s) &= \sum_i \sum_j s[i, j] f(\delta_{i, j}) \\
&= \sum_i \sum_j s[i, j] f \circ t_{i, j}(\delta_{0, 0})
\end{aligned}$$



$$= \sum_i \sum_j s[i, j] t_{i,j} \circ f(\delta_{0,0})$$

126 By denoting  $w = f(\delta_{0,0}) \in \mathcal{S}(\mathbb{Z}^2)$ , we obtain:

$$\begin{aligned} \forall (a, b) \in \mathbb{Z}^2, f(s)[a, b] &= \sum_i \sum_j s[i, j] t_{i,j}(w)[a, b] \\ &= \sum_i \sum_j s[i, j] w[a - i, b - j] \\ \text{i.e. } f(s) &= s * w \end{aligned} \tag{2}$$

127

□

### 128 2.1.3 Usefulness of convolutions in deep learning

#### 129 Equivariance property of CNNs

130 In deep learning, an important argument in favor of CNNs is that convolu-  
131 tional layers are equivariant to translations. Intuitively, that means that a  
132 detail of an object in an image should produce the same features indepen-  
133 dently of its position in the image.

#### 134 Lossless superiority of CNNs over MLPs

135 The converse result, as a consequence of Proposition 6, is never mentioned  
136 in deep learning literature. However it is also a strong one. For example,  
137 let's consider a linear function that is equivariant to translations. Thanks  
138 to the converse result, we know that this function is a convolution operator  
139 parameterized by a weight vector  $w$ ,  $f_w : \cdot * w$ . If the domain is compactly  
140 supported, as in the case of images, we can break down the information of  $w$   
141 in a finite number  $n_q$  of kernels  $w_q$  with small compact supports of same size  
142 (for instance of size  $2 \times 2$ ), such that we have  $f_w = \sum_{q \in \{1, 2, \dots, n_q\}} f_{w_q}$ . The  
143 convolution operators  $f_{w_q}$  are all in the search space of  $2 \times 2$  convolutional  
144 layers. In other words, every translational equivariant linear function can

145 have its information parameterized by these layers. So that means that the  
146 reduction of parameters from an MLP to a CNN is done with strictly no loss of  
147 expressivity (provided the objective function is known to bear this property).  
148 Besides, it also helps the training to search in a much more confined space.

149 **Methodology for extending to general graphs**

150 Hence, in our construction, we will try to preserve the characterization from  
151 Proposition 6 as it is mostly the reason why they are successful in deep  
152 learning. Note that the reduction of parameters compared to a dense layer  
153 is also a consequence of this characterization.

## 2.2 Construction from the vertex set

As Proposition 6 is a complete characterization of convolutions, it can be used to define them *i.e.* convolution operators can be constructed as the set of linear transformations that are equivariant to translations. However, in the general case where  $G$  is not a grid graph, translations are not defined, so that construction needs to be generalized beyond translational equivariances. In mathematics, convolutions are more generally defined for signals defined over a group structure. The classical convolution that is used in deep learning is just a narrow case where the domain group is an euclidean space. Therefore, constructing a convolution on graphs should start from the more general definition of convolution on groups rather than convolution on euclidean domains.

Our construction is motivated by the following questions:

- Does the equivariance property holds ? Does the characterization from Proposition 6 still holds ?
- Is it possible to extend the construction on non-group domains, or at least on mixed domains ? (*i.e.* one signal is defined over a set, and the other is defined over a subgroup of the transformations of this set).
- Can a group domain draw an underlying graph structure ? Is the group convolution naturally defined on this class of graphs ?

We first recall the notion of group and group convolution.

### Definition 7. Group

A group  $\Gamma$  is a set equipped with a closed, associative and invertible composition law that admits a unique left-right identity element.

The group convolution extends the notion of the classical discrete convolution.

180 **Definition 8. Group convolution I**

181 Let a group  $\Gamma$ , the group convolution I between two signals  $s_1$  and  $s_2 \in \mathcal{S}(\Gamma)$   
 182 is defined as:

$$\forall h \in \Gamma, (s_1 *_I s_2)[h] = \sum_{g \in \Gamma} s_1[g] s_2[g^{-1}h]$$

183 provided at least one of the signals has finite support if  $\Gamma$  is not finite.

184 **2.2.1 Steered construction from groups**

185 For a graph  $G = \langle V, E \rangle$  and a subgroup  $\Gamma \subset \Phi^*(V)$  or its invertible transfor-  
 186 mations, Definition 8 is applicable for  $\mathcal{S}(\Gamma)$ , but not for  $\mathcal{S}(V)$  as  $V$  is not a  
 187 group. Nonetheless, our point here is that we will use the group convolution  
 188 on  $\mathcal{S}(\Gamma)$  to construct the convolutions on  $\mathcal{S}(V)$ .

189 For now, let's assume  $\Gamma$  is in one-to-one correspondence with  $V$ , and let's  
 190 define a bijective map  $\varphi$  from  $\Gamma$  to  $V$ . We denote  $\Gamma \xrightarrow{\varphi} V$  and  $g_v \xrightarrow{\varphi} v$ .

191 Then, the linear morphism  $\tilde{\varphi}$  from  $\mathcal{S}(\Gamma)$  to  $\mathcal{S}(V)$  defined on the Dirac bases  
 192 by  $\tilde{\varphi}(\delta_g) = \delta_{\varphi(g)}$  is a linear isomorphism. Hence,  $\mathcal{S}(V)$  would inherit the same  
 193 inherent structural properties as  $\mathcal{S}(\Gamma)$ . For the sake of notational simplicity,  
 194 we will use the same symbol  $\varphi$  for both  $\varphi$  and  $\tilde{\varphi}$  (as done between  $f$  and  
 195  $L_f$ ). A commutative diagram between the sets is depicted on Figure 2.1.

$$\begin{array}{ccc} \Gamma & \xrightarrow{\varphi} & V \\ s \downarrow & & \downarrow s \\ \mathcal{S}(\Gamma) & \xrightarrow{\varphi} & \mathcal{S}(V) \end{array}$$

Figure 2.1: Commutative diagram between sets

196 We naturally obtain the following relation, which put in simpler words means  
 197 that signals on  $\mathcal{S}(\Gamma)$  are mapped to  $\mathcal{S}(V)$  when  $\varphi$  is simultaneously applied  
 198 on both the signal space and its domain.

199 **Lemma 9. Relation between  $\mathcal{S}(\Gamma)$  and  $\mathcal{S}(V)$**

200  $\forall s \in \mathcal{S}(\Gamma), \forall u \in V, \varphi(s)[u] = s[\varphi^{-1}(u)] = s[g_u]$

*Proof.*

$$\begin{aligned} \forall s \in \mathcal{S}(\Gamma), \varphi(s) &= \varphi\left(\sum_{g \in \Gamma} s[g] \delta_g\right) = \sum_{g \in \Gamma} s[g] \varphi(\delta_g) = \sum_{g \in \Gamma} s[g] \delta_{\varphi(g)} \\ &= \sum_{v \in V} s[g_v] \delta_v \end{aligned}$$

So  $\forall v \in V, \varphi(s)[u] = s[g_u]$

201

□

202 Hence, we can steer the definition of the group convolution from  $\mathcal{S}(\Gamma)$  to  
203  $\mathcal{S}(V)$  as follows:

204 **Definition 10. Group convolution II**

205 Let a subgroup  $\Gamma \subset \Phi^*(V)$  such that  $\Gamma \stackrel{\varphi}{\cong} V$ . The group convolution II  
206 between two signals  $s_1$  and  $s_2 \in \mathcal{S}(V)$  is defined as:

$$\forall u \in V, (s_1 *_\text{II} s_2)[u] = \sum_{v \in V} s_1[v] s_2[\varphi(g_v^{-1} g_u)]$$

207

208 **Lemma 11. Relation between group convolution I and II**

209 Let a subgroup  $\Gamma \subset \Phi^*(V)$  such that  $\Gamma \stackrel{\varphi}{\cong} V$ ,

$$\forall s_1, s_2 \in \mathcal{S}(\Gamma), \forall u \in V, (\varphi(s_1) *_\text{II} \varphi(s_2))[u] = (s_1 *_\text{I} s_2)[g_u]$$

210

211 *Proof.* Using Lemma 9,

$$\begin{aligned}
 (\varphi(s_1) *_{\text{II}} \varphi(s_2))[u] &= \sum_{v \in V} \varphi(s_1)[v] \varphi(s_2)[\varphi(g_v^{-1} g_u)] \\
 &= \sum_{v \in V} s_1[g_v] s_2[g_v^{-1} g_u] \\
 &= \sum_{g \in \Gamma} s_1[g] s_2[g^{-1} g_u] \\
 &= (s_1 *_{\text{I}} s_2)[g_u]
 \end{aligned}$$

212

□

213 For convolution II, we only obtain a weak version of Proposition 6.

214 **Proposition 12. Equivariance to  $\varphi(\Gamma)$**

215 If  $\varphi$  is a homomorphism, convolution operators acting on the right of  $\mathcal{S}(V)$   
 216 are equivariant to  $\varphi(\Gamma)$  i.e.

$$\begin{aligned}
 &\text{if } \varphi \in \text{ISO}(\Gamma, V), \\
 &\exists w \in \mathcal{S}(V), f = . *_{\text{II}} w \Rightarrow \forall v \in V, f \circ \varphi(g_v) = \varphi(g_v) \circ f
 \end{aligned}$$

217

*Proof.*

$$\begin{aligned}
 &\forall s \in \mathcal{S}(V), \forall u \in V, \forall v \in V, \\
 (f_w \circ \varphi(g_u))(s)[v] &= \sum_{v \in V} \varphi(g_u)(s)[v] w[\varphi(g_v^{-1} g_u)] \\
 &= \sum_{\substack{(a,b) \in V^2 \\ \text{s.t. } g_a g_b = g_v}} \varphi(g_u)(s)[a] w[b] \\
 &= \sum_{\substack{(a,b) \in V^2 \\ \text{s.t. } g_a g_b = g_v}} s[\varphi(g_u)^{-1}(a)] w[b]
 \end{aligned}$$

$$= \sum_{\substack{(a,b) \in V^2 \\ s.t. \ g_{\varphi(g_u)(a)} g_b = g_v}} s[a] w[b]$$

218 Because  $\varphi$  is an isomorphism, its inverse  $c \mapsto g_c$  is also an isomorphism and

219 so  $g_{\varphi(g_u)(a)} g_b = g_v \Leftrightarrow g_a g_b = g_{\varphi(g_u)^{-1}(v)}$ . So we have both:

$$\begin{aligned} (f_w \circ \varphi(g_u))(s)[v] &= \sum_{\substack{(a,b) \in V^2 \\ s.t. \ g_a g_b = g_{\varphi(g_u)^{-1}(v)}}} s[a] w[b] \\ &= s *_\Pi w[\varphi(g_u)^{-1}(v)] \\ &= (\varphi(g_u) \circ f_w)(s)[v] \end{aligned}$$

220

□

221 *Remark.* Note that convolution operators of the form  $f_w = . *_\Pi w$  are also  
222 equivariant to  $\Gamma$ , but the proposition and the proof are omitted as they are  
223 similar to the latter.

224 In fact, both group convolutions are the same as the latter one borrows the  
225 algebraic structure of the first one. Thus we only obtain equivariance to  $\varphi(\Gamma)$   
226 when  $\varphi$  also transfer the group structure from  $\Gamma$  to  $V$ , and the converse don't  
227 hold. To obtain equivariance to  $\Gamma$  (and its converse), we will drop the direct  
228 homomorphism condition, and instead we will take into account the fact that  
229 it contains invertible transformations of  $V$ .

### 230 2.2.2 Construction under group actions

231 **Definition 13. Group action**

232 An *action* of a group  $\Gamma$  on a set  $V$ , is a function  $L : \Gamma \times V \rightarrow V, (g, v) \mapsto L_g(v)$ ,  
 233 such that the map  $g \mapsto L_g$  is a homomorphism.

234 Given  $g \in \Gamma$ , the transformation  $L_g$  is called the action of  $g$  by  $L$  on  $V$ .

235 *Remark.* When there is no ambiguity, we use the same symbol for  $g$  and  $L_g$ .

236 Hence, note that  $g \in \Gamma$  can act on both  $\Gamma$  through the left multiplication  
 237 and on  $V$  as being an object of  $\Phi^*(V)$ . This ambivalence can be seen on a  
 238 commutative diagram, see Figure 2.2.

$$\begin{array}{ccc} g_u & \xrightarrow{g_v} & g_v g_u \\ \varphi \downarrow & & \downarrow \varphi \\ u & \xrightarrow[g_v]{(P)} & \varphi(g_v g_u) \end{array}$$

Figure 2.2: Commutative diagram. All arrows except for the one labeled with (P) are always True.

239 For (P) to be true means that  $\varphi$  is an equivariant map *i.e.* whether the  
 240 mapping is done before or after the action of  $\Gamma$  has no impact on the result.  
 241 When such  $\varphi$  exists,  $\Gamma$  and  $V$  are said to be equivalent and we denote  $\Gamma \equiv V$ .

242 **Definition 14. Equivariant map**

243 A map  $\varphi$  from a group  $\Gamma$  acting on the destination set  $V$  is said to be an  
 244 *equivariant map* if

$$\forall g, h \in \Gamma, g(\varphi(h)) = \varphi(gh)$$

245

246 In our case we have  $\Gamma \stackrel{\varphi}{\cong} V$ . If we also have that  $\Gamma \equiv V$ , we are interested to  
 247 know if then  $\varphi$  exhibits the equivalence.



248 **Definition 15.  $\varphi$ -Equivalence**

249 A subgroup  $\Gamma \subset \Phi^*(V)$  such that  $\Gamma \stackrel{\varphi}{\cong} V$ , is said to be  $\varphi$ -equivalent if  $\varphi$  is a  
 250 bijective equivariant map *i.e.* if it verifies the property:

$$\forall v, u \in V, g_v(u) = \varphi(g_v g_u) \quad (\text{P})$$

251 In that case we denote  $\Gamma \stackrel{\varphi}{\equiv} V$ .

252 *Remark.* For example, translations on the grid graph, with  $\varphi(t_{i,j}) = (i, j)$ ,  
 253 are  $\varphi$ -equivalent as  $t_{i,j}(a, b) = \varphi(t_{i,j} \circ t_{a,b})$ . However, with  $\varphi(t_{i,j}) = (-i, -j)$ ,  
 254 they would not be  $\varphi$ -equivalent.

255 **Definition 16. Group convolution III**

256 Let a subgroup  $\Gamma \subset \Phi^*(V)$  such that  $\Gamma \stackrel{\varphi}{\cong} V$ . The group convolution III  
 257 between two signals  $s_1$  and  $s_2 \in \mathcal{S}(V)$  is defined as:

$$s_1 *_{\text{III}} s_2 = \sum_{v \in V} s_1[v] g_v(s_2) \quad (3)$$

$$= \sum_{g \in \Gamma} s_1[\varphi(g)] g(s_2) \quad (4)$$

258

259 The two expressions differ on the domain upon which the summation is done.  
 260 The expression (3) put the emphasis on each vertex and its action, whereas  
 261 the expression (4) emphasizes on each object of  $\Gamma$ .

262 **Lemma 17. Relation with group convolution II**

263  $\Gamma \stackrel{\varphi}{\equiv} V \Leftrightarrow *_{\text{II}} = *_{\text{III}}$

*Proof.*

$$\forall s_1, s_2 \in \mathcal{S}(V),$$

$$\begin{aligned} s_1 *_{\text{II}} s_2 &= s_1 *_{\text{III}} s_2 \\ \Leftrightarrow \forall u \in V, \sum_{v \in V} s_1[v] s_2[\varphi(g_v^{-1} g_u)] &= \sum_{v \in V} s_1[v] s_2[g_v^{-1}(u)] \end{aligned} \quad (5)$$

264 Hence, the direct sense is obtained by applying (P).  
 265 For the converse, given  $u, v \in V$ , we first realize (5) for  $s_1 := \delta_v$ , obtaining  
 266  $s_2[\varphi(g_v^{-1}g_u)] = s_2[g_v^{-1}(u)]$ , which we then realize for a real signal  $s_2$  having no  
 267 two equal entries, obtaining  $\varphi(g_v^{-1}g_u) = g_v^{-1}(u)$ . From the latter we finally  
 268 obtain (P) with the one-to-one correspondence  $g_{v'} := g_v^{-1}$ .  $\square$

269 We can then coin the term  $\varphi$ -convolution.

270 **Definition 18.  $\varphi$ -convolution**

271 Let  $\Gamma \stackrel{\varphi}{\cong} V$ , the  $\varphi$ -convolution between two signals  $s_1$  and  $s_2 \in \mathcal{S}(V)$  is  
 272 defined as:

$$s_1 *_{\varphi} s_2 = s_1 *_{\text{II}} s_2 = s_1 *_{\text{III}} s_2$$

273

274 This time, we do obtain equivariance to  $\Gamma$  as expected, and the full charac-  
 275 terization as well.

276 **Proposition 19. Characterization by right-action equivariance to  $\Gamma$**

277 If  $\Gamma$  is  $\varphi$ -equivalent, the class of linear transformations of  $\mathcal{S}(V)$  that are  
 278 equivariant to  $\Gamma$  is exactly the class of  $\varphi$ -convolution operators acting on the  
 279 right of  $\mathcal{S}(V)$  *i.e.*

$$\begin{aligned} &\text{If } \Gamma \stackrel{\varphi}{\cong} V, \\ &\exists w \in \mathcal{S}(V), f = . *_{\varphi} w \Leftrightarrow \begin{cases} f \in \mathcal{L}(\mathcal{S}(V)) \\ \forall g \in \Gamma, f \circ g = g \circ f \end{cases} \end{aligned}$$

280

281 *Proof.* 1. From left to right:

282 In the following equations, (6) is obtained by definition, (7) is obtained  
 283 because left multiplication in a group is bijective, and (8) is obtained

284 because of (P).

$$\forall g \in \Gamma, \forall s \in \mathcal{S}(V),$$

$$f_w \circ g(s) = \sum_{h \in \Gamma} g(s)[\varphi(h)] h(w) \quad (6)$$

$$= \sum_{h \in \Gamma} g(s)[\varphi(gh)] gh(w) \quad (7)$$

$$= \sum_{h \in \Gamma} g(s)[g(\varphi(h))] gh(w) \quad (8)$$

$$= \sum_{h \in \Gamma} s[\varphi(h)] gh(w)$$

$$= \sum_{h \in \Gamma} s[\varphi(h)] h(w)[g^{-1}(.)]$$

$$= f_w(s)[g^{-1}(.)]$$

$$= g \circ f_w(s)$$

285 Of course, we also have that  $f_w$  is linear.

286 2. From right to left:

287 Let  $f \in \mathcal{L}(\mathcal{S}(V))$ ,  $s \in \mathcal{S}(V)$ . By linearity of  $f$ , we distribute  $f(s)$  on  
288 the family of dirac signals:

$$f(s) = \sum_{v \in V} s[v] f(\delta_v) \quad (9)$$

289 Thanks to (P), we have that:

$$g_v(\varphi(\text{Id})) = \varphi(g_v \text{Id}) = v$$

$$\text{So, } v = u \Leftrightarrow \varphi(\text{Id}) = g_v^{-1}(u)$$

$$\text{So, } \delta_v = g_v(\delta_{\varphi(\text{Id})})$$

290 By denoting  $w = f(\delta_{\varphi(\text{Id})})$ , and using the hypothesis of equivariance,

we obtain from (9) that:

$$\begin{aligned}
 f(s) &= \sum_{v \in V} s[v] f \circ g_v(\delta_{\varphi(\text{Id})}) \\
 &= \sum_{v \in V} s[v] g_v \circ f(\delta_{\varphi(\text{Id})}) \\
 &= \sum_{v \in V} s[v] g_v(w) \\
 &= s *_{\varphi} w
 \end{aligned}$$

□

### Construction of $\varphi$ -convolutions on vertex domains

Proposition 19 tells us that in order to define a convolution on the vertex domain of a graph  $G = \langle V, E \rangle$ , all we need is a subgroup  $\Gamma$  of invertible transformations of  $V$ , that is equivalent to  $V$ . The choice of  $\Gamma$  can be done with respect to  $E$ . This is discussed in more details in Section 2.3, where we will see that in fact, we only need a generating set of  $\Gamma$ .

### Exposure of $\varphi$

This construction relies on exposing a bijective equivariant map  $\varphi$  between  $\Gamma$  and  $V$ . In the next subsection, we show that in cases where  $\Gamma$  is abelian, we even need not expose  $\varphi$  and the characterization still holds.

### 2.2.3 Mixed domain formulation

From (4), we can define a mixed domain convolution *i.e.* that is defined for  $r \in \mathcal{S}(\Gamma)$  and  $s \in \mathcal{S}(V)$ , without the need of expliciting  $\varphi$ .

**Definition 20. Mixed domain convolution**

Let a subgroup  $\Gamma \subset \Phi^*(V)$  such that  $V \cong \Gamma$ . The *mixed domain convolution* between two signals  $r \in \mathcal{S}(\Gamma)$  and  $s \in \mathcal{S}(V)$  results in a signal  $r *_{\text{M}} s \in \mathcal{S}(V)$  and is defined as:

$$r *_{\text{M}} s = \sum_{g \in \Gamma} r[g] g(s)$$

We coin it M-convolution. From a practical point of view, this expression of the convolution is useful because it relegates  $\varphi$  as an underpinning object.

**Lemma 21. Relation with group convolution III**

$\forall \varphi \in \text{BIJ}(\Gamma, V), \forall (r, s) \in \mathcal{S}(\Gamma) \times \mathcal{S}(V),$

$$r *_{\text{M}} s = \varphi(r) *_{\text{III}} s$$

*Proof.* Let  $\varphi \in \text{BIJ}(\Gamma, V), (r, s) \in \mathcal{S}(\Gamma) \times \mathcal{S}(V),$

$$\begin{aligned} r *_{\text{M}} s &= \sum_{g \in \Gamma} r[g] g(s) = \sum_{v \in V} r[g_v] g_v(s) \stackrel{(\diamond)}{=} \sum_{v \in V} \varphi(r)[v] g_v(s) \\ &= \varphi(r) *_{\text{III}} s \end{aligned}$$

Where  $\stackrel{(\diamond)}{=}$  comes from Lemma 9. □

In other words,  $*_{\text{M}}$  is a convenient reformulation of  $*_{\text{III}}$  which does not depend on a particular  $\varphi$ .

**Lemma 22. Relation with group convolution I, II and  $\varphi$ -convolution**

Let  $\varphi \in \text{BIJ}(\Gamma, V), (r, s) \in \mathcal{S}(\Gamma) \times \mathcal{S}(V),$  we have:

$$\begin{aligned} \Gamma \stackrel{\varphi}{=} V &\Leftrightarrow \forall v \in V, (r *_{\text{M}} s)[v] = (r *_{\text{I}} \varphi^{-1}(s))[g_v] \\ &\Leftrightarrow r *_{\text{M}} s = \varphi(r) *_{\text{II}} s \\ &\Leftrightarrow r *_{\text{M}} s = \varphi(r) *_{\varphi} s \end{aligned}$$

323

324 *Proof.* On one hand, Lemma 21 gives  $r *_M s = \varphi(r) *_{III} s$ . On the other hand,  
 325 Lemma 11 gives  $\forall v \in V, (r *_I \varphi^{-1}(s))[g_v] = (\varphi(r) *_{II} s)[v]$ . Then Lemma 17  
 326 concludes.  $\square$

327 *Remark.* The converse sense is meaningful because it justifies that when the  
 328 M-convolution is employed, the property  $\Gamma \equiv V$  underlies, without the need  
 329 of expliciting  $\varphi$ .

330 From M-convolution, we can derive operators acting on the left of  $\mathcal{S}(V)$ , of  
 331 the form  $s \mapsto w *_M s$ , parameterized by  $w \in \mathcal{S}(\Gamma)$ . In particular, these  
 332 operators would be relevant as layers of neural networks. On the contrary,  
 333 derived operators acting on the right such as  $r \mapsto r *_M w$  wouldn't make  
 334 sense with this formulation as they would make  $\varphi$  resurface. However, the  
 335 equivariance to  $\Gamma$  incurring from Lemma 21 and Proposition 19 only holds for  
 336 operators acting on the right. So we need to intertwine an abelian condition  
 337 as follows. This is also a good excuse to see the influence of abelianity.

### 338 **Proposition 23. Equivariance to $\Gamma$ through left action**

339 Let a subgroup  $\Gamma \subset \Phi^*(V)$  such that  $\Gamma \cong V$ .  $\Gamma$  is abelian, if and only if,  
 340 M-convolution operators acting on the left of  $\mathcal{S}(V)$  are equivariant to it *i.e.*

$$\forall g, h \in \Gamma, gh = hg \Leftrightarrow \forall w, g \in \Gamma, w *_M g(.) = g \circ (w *_M .)$$

341 *Proof.* Let  $w, g \in \Gamma$ , and define  $f_w : s \mapsto w *_M s$ . In the following expressions,  
 342  $\Gamma$  is abelian if and only if (10) and (11) are equal (the converse is obtained

343 by particularizing on well chosen signals):

$$f_w \circ g(s) = \sum_{h \in \Gamma} w[h] hg(s) \quad (10)$$

$$= \sum_{h \in \Gamma} w[h] gh(s) \quad (11)$$

$$= \sum_{h \in \Gamma} w[h] h(s)[g^{-1}(.)]$$

$$= (w *_{\mathbf{M}} s)[g^{-1}(.)]$$

$$= g \circ f_w(s)$$

344

□

345 *Remark.* Similarly,  $*_{\varphi}$  is also equivariant to  $\Gamma$  through left action if and only  
 346 if  $\Gamma$  is abelian, as a consequence of being commutative if and only if  $\Gamma$  is  
 347 abelian. On the contrary, note that commutativity of  $*_{\mathbf{M}}$  doesn't make sense.

348 **Corrolary 24. Characterization by left-action equivariance to  $\Gamma$**

349 Let  $\Gamma \cong V$ . If  $\Gamma$  is abelian, the class of linear transformations of  $\mathcal{S}(V)$  that  
 350 are equivariant to  $\Gamma$  is exactly the class of M-convolution operators acting on  
 351 the left of  $\mathcal{S}(V)$  *i.e.*

If  $\Gamma \cong V$  and  $\Gamma$  is abelian,

$$\exists w \in \mathcal{S}(\Gamma), f = w *_{\mathbf{M}} . \Leftrightarrow \begin{cases} f \in \mathcal{L}(\mathcal{S}(V)) \\ \forall g \in \Gamma, f \circ g = g \circ f \end{cases}$$

352

353 *Proof.* By picking  $\varphi$  such that  $\Gamma \stackrel{\varphi}{\cong} V$  with Lemma 22 and using the relation  
 354 between  $*_{\mathbf{M}}$  and  $*_{\varphi}$ . □

355 Depending on the applications, we will build upon either  $*_{\varphi}$  or  $*_{\mathbf{M}}$  when the  
 356 abelian condition is satisfied.

## 2.3 Inclusion of the edge set in the construction

The constructions from the previous section involve the vertex set  $V$  and depend on  $\Gamma$ , a subgroup of the set of invertible transformations on  $V$ . Therefore, it looks natural to try to relate the edge set and  $\Gamma$ .

There are two approaches. Either  $\Gamma$  describes an underlying graph structure  $G = \langle V, E \rangle$ , either  $G$  can be used to define a relevant subgroup  $\Gamma$  to which the produced convolutive operators will be equivariant. Both approaches will help characterize classes of graphs that can support natural definitions of convolutions.

TODO: introduction to be developped more

### 2.3.1 Convolution on Cayley subgraphs

In this subsection, we are trying to answer the following question:

- What graphs admit a  $\varphi$ -convolution, or an M-convolution (in the sense that they can be defined with the characterization), under the condition that  $\Gamma$  is generated by a set of edge-constrained transformations ?

#### Definition 25. Edge-constrained transformation

An *edge-constrained* (EC) transformation on a graph  $G = \langle V, E \rangle$  is a transformation  $f : V \mapsto V$  such that

$$\forall u, v \in V, f(u) = v \Rightarrow u \stackrel{E}{\sim} v$$

We denote  $\Phi_{\text{EC}}(G)$  and  $\Phi_{\text{EC}}^*(G)$  the sets of (EC) and invertible (EC) transformations. When a convolution is defined as a sum over a set that is in one-to-one correspondence with a group that is generated from a set of (EC) transformations, we call it an (EC) convolution.



380 *Remark.* Note that  $\Phi_{\text{EC}}^*(G)$  is not a group, thus why we are interested in  
 381 groups and their generating sets.

382 This leads us to consider Cayley graphs (Cayley, 1878; Wikipedia, 2018).

383 **Definition 26. Cayley graph**

384 Let a group  $\Gamma$  and one of its generating set  $\mathcal{U}$ . The *Cayley graph* generated  
 385 by  $\mathcal{U}$ , is the digraph  $\vec{G} = \langle V, E \rangle$  such that  $V = \Gamma$  and  $E$  is such that:

$$u \rightarrow v \Leftrightarrow \exists g \in \mathcal{U}, ga = b$$

386 Also, if  $\Gamma$  is abelian, we call it an *abelian Cayley graph*. We call *Cayley*  
 387 *subgraph*, a subgraph that is isomorph to a Cayley graph.

388 *Remark.* Note that for compatibility with the functional notation that we  
 389 use, we define Cayley graphs with  $ga = b$  instead of  $ag = b$ .

390 **Convolution on Cayley graphs**

391 In the case of Cayley graphs, it is clear that  $\mathcal{U} \subseteq \Phi_{\text{EC}}^*$  and  $\Phi^* \supseteq \langle \mathcal{U} \rangle \equiv V$ .  
 392 So that they admit (EC)  $\varphi$ -convolutions, and (EC) M-convolutions in the  
 393 abelian case.

394 More precisely, we obtain the following characterization:

395 **Proposition 27. Characterization by Cayley subgraph isomorphism**

396 Let a graph  $G = \langle V, E \rangle$ , then:

397 (i)  $G$  admits an (EC)  $\varphi$ -convolution if and only if it contains a subgraph  
 398 isomorph to a Cayley graph

399 (ii)  $G$  admits an (EC) M-convolution if and only if it contains a subgraph  
 400 isomorph to an abelian Cayley graph

401 *Proof.* We show the result only in the general case as the proof for the abelian  
 402 case is similar.

403 1. From left to right: as a direct application of the definitions.

404 2. From right to left:

405 Let a graph  $G = \langle V, E \rangle$ . We suppose it contains a subgraph  $\vec{G}_s =$   
 406  $\langle V_s, E_s \rangle$  that is graph-isomorph to a Cayley graph  $\vec{G}_c = \langle V_c, E_c \rangle$ , gen-  
 407 erated by  $\mathcal{U}$ . Let  $\psi$  be a graph isomorphism from  $G_s$  to  $G_c$ . To obtain  
 408 the proof, we need to find a group of invertible transformations  $\Gamma$  of  $V_s$   
 409 generated by a set of (EC) transformations, such that  $\Gamma \equiv V_s$ .

410 Let's define the group action  $L : V_c \times V_s \rightarrow V_s$  inductively as follows:

411 (a)  $\forall g \in \mathcal{U}, L_g(u) = v \Leftrightarrow g\psi(u) = \psi(v)$

412 (b) Whenever  $L_g$  and  $L_h$  are defined, the action of  $gh$  is defined by  
 413 homomorphism as  $L_{gh} = L_g \circ L_h$

414 (c) Whenever  $L_g$  is defined, the action of  $g^{-1}$  is defined by homomor-  
 415 phism as  $L_{g^{-1}} = L_g^{-1}$  *i.e.*  $L_{g^{-1}}(u) = v \Leftrightarrow \psi(u) = g\psi(v)$

416 Note that the induction transfers the property (a) to all  $g \in V_c$  in a  
 417 transitive manner because

$$L_{gh}(u) = L_g(L_h(u)) = w \Leftrightarrow \exists v \in V_s \begin{cases} L_h(u) = v \\ L_g(v) = w \end{cases}$$

418 and

$$\exists v \in V_s \begin{cases} h\psi(u) = \psi(v) \\ g\psi(v) = \psi(w) \end{cases} \Leftrightarrow gh\psi(u) = \psi(w)$$

419 We must also verify that this construction is well-defined, *i.e.* whenever  
 420 we define an action with (b) or (c), if the action was already defined,  
 421 then they must be equal. This is the case because the homomorphism

422  $g \mapsto L_g$  on  $V_c$  is in fact an isomorphism as

$$\begin{aligned} L_g = L_h &\Leftrightarrow \forall u \in V, L_g(u) = L_h(u) \\ &\Leftrightarrow \forall u \in V, g\psi(u) = h\psi(u) \\ &\Leftrightarrow g = h \end{aligned}$$

423 Also note that (c) is needed only in case that  $V_c$  is infinite.

424 Denote the set  $L_{\mathcal{U}} = \{L_g, g \in \mathcal{U}\}$  and  $\Gamma = \langle L_{\mathcal{U}} \rangle \cong V_c$ . Let's define the  
425 map  $\varphi$  as:

$$\begin{aligned} \Gamma &\rightarrow V_s \\ \varphi : L_g &\mapsto L_g(\psi^{-1}(\text{Id})) \end{aligned}$$

426  $\varphi$  is bijective because  $\forall g \in V_c, \varphi(L_g) = \psi^{-1}(g)$  thanks to (a).

427 Additionally, we have:

$$\begin{aligned} L_h(\varphi(L_g)) &= L_h(L_g(\psi^{-1}(\text{Id}))) \\ &= L_h \circ L_g(\psi^{-1}(\text{Id})) \\ &= L_{hg}(\psi^{-1}(\text{Id})) \\ &= \varphi(L_{hg}) \\ &= \varphi(L_h \circ L_g) \end{aligned}$$

428 That is,  $\varphi$  is a bijective equivariant map and  $\langle L_{\mathcal{U}} \rangle = \Gamma \stackrel{\varphi}{\cong} V_s$ . Moreover,  
429  $L_{\mathcal{U}}$  is a set of (EC) transformations thanks to (a). Therefore,  $G$  admits  
430 an (EC)  $\varphi$ -convolution.

431

□

432 **Corrolary 28. Characterization by  $\varphi$**

433 Let a graph  $G = \langle V, E \rangle$ , and a set  $\mathcal{U} \subset \Phi_{\text{EC}}^*(G)$  s.t.

$$\langle \mathcal{U} \rangle \cong \Gamma \equiv V' \subset V$$

434  $G$  admits an (EC)  $\varphi$ -convolution, if and only if,  $\varphi$  is a graph isomorphism  
435 between the Cayley graph generated by  $\mathcal{U}$  and the subgraph induced by  $V'$ .

436 The proof is omitted as it would be highly similar to the previous one.

### 437 2.3.2 Intrinsic properties

438 • Obviously the contructed convolutions are linear. But do they also  
439 preserve the locality and weight sharing properties ?

440 Let  $\vec{G} = \langle V, E \rangle$  be a Cayley subgraph, generated by  $\mathcal{U}$ , of some graph  $G$ .  
441 Recall that its (EC)  $\varphi$ -convolution operator is a right operator, and can be  
442 expressed as

$$\begin{aligned} \forall s \in \mathcal{S}(V), \forall u \in V, \\ f_w(s)[u] &= (s *_{\varphi} w)[u] \\ &= \sum_{v \in V} s[v] w[g_v^{-1}(u)] \end{aligned} \tag{12}$$

443 From this expression, it is not obvious that  $f_w$  is a local operator. To see  
444 this, we can show for example the following proposition.

### 445 **Proposition 29. Locality**

446 When the support of  $w$  is a compact (in the sense that its induced subgraph  
447 in  $G$  is connected), of diameter  $d$ , the same holds for the support of the  
448 sum  $\Sigma$  in (12). More precisely, the subgraph induced by the support of  $\Sigma$  is  
449 isomorphic to the transpose of the subgraph induced by the support of  $w$ .

450 *Proof.* Without loss of generality subject to growing  $\mathcal{U}$ , let's suppose that  
 451  $w$  has a support  $\mathcal{M} = \varphi(\mathcal{N})$ , such that  $\mathcal{N} \subset \mathcal{U}$ .  $\mathcal{N}$  and  $\mathcal{M}$  are obviously  
 452 compacts of diameter 2. Thanks to (P), we have

$$\begin{aligned}
 g_v^{-1}(u) \in \mathcal{M} &\Leftrightarrow u \in g_v(\mathcal{M}) = g_v(\varphi(\mathcal{N})) = \varphi(g_v\mathcal{N}) \\
 &\Leftrightarrow g_u \in g_v\mathcal{N} \\
 &\Leftrightarrow g_v^{-1} \in \mathcal{N}g_u^{-1} \\
 &\Leftrightarrow g_v \in g_u\mathcal{N}^{-1} \\
 &\Leftrightarrow v \in g_u(\varphi(\mathcal{N}^{-1}))
 \end{aligned}$$

453 where  $\mathcal{N}^{-1}$  reverses the edges of  $\mathcal{N}$ . Let's denote  $\mathcal{K}_u = g_u(\varphi(\mathcal{N}^{-1})) \subset V$ .  
 454 By composing edge reversal and graph isomorphisms (as  $\varphi$  and its inverse are  
 455 graph isomorphisms by Proposition 28), the compactness and diameter of  $\mathcal{M}$   
 456 is preserved for  $\mathcal{K}_u$ . More preceisely, the transposed subgraph structure is  
 457 also preserved.  $\square$

458 Let's define  $\mathcal{M}$ ,  $\mathcal{N}$  and  $\mathcal{K}_u$  as in the previous proof.

459 **Definition 30. Supporting set**

460 The *supporting set* of an (EC) convolution operator  $f_w$ , is a set  $\mathcal{N} \subset \Phi_{\text{EC}}^*$ ,  
 461 such that

462 (i) when  $*$  is  $*_{\varphi}$ :  $0 \notin w[\mathcal{M}]$ , where  $\mathcal{M} = \varphi(\mathcal{N})$

463 (ii) when  $*$  is  $*_{\text{M}}$ :  $0 \notin w[\mathcal{N}]$

464 **Definition 31. Local patch for  $*_{\varphi}$**

465 The *local patch* at  $u \in V$  of an (EC)  $\varphi$ -convolution operator  $f_w$  is defined as  
 466  $\mathcal{K}_u = g_u(\varphi(\mathcal{N}^{-1}))$ .

467 To see that the weights are tied in the general case (i), we can show the  
 468 following proposition.

469 **Proposition 32. Weight sharing**

470  $\forall a, \alpha \in V, \forall b \in \mathcal{K}_a : \exists \beta \in \mathcal{K}_\alpha \Leftrightarrow g_\beta^{-1}(\alpha) = g_b^{-1}(a)$

471 *Proof.* By using (P),

$$\begin{aligned} g_{\mathcal{K}_\alpha}^{-1}(\alpha) = g_{\mathcal{K}_a}^{-1}(a) &\Leftrightarrow g_\alpha^{-1}g_{\mathcal{K}_\alpha} = g_a^{-1}g_{\mathcal{K}_a} \\ &\Leftrightarrow \mathcal{K}_\alpha = g_\alpha g_a^{-1}(\mathcal{K}_a) = g_\alpha g_a^{-1}g_a(\varphi(\mathcal{N}^{-1})) \\ &\Leftrightarrow \mathcal{K}_\alpha = g_\alpha(\varphi(\mathcal{N}^{-1})) \end{aligned}$$

472

□

### 473 2.3.3 Stricly edge-constrained convolutions

474 We make the disctinction between general (EC) convolution operators and  
475 those for which the weight kernel  $w$  is smaller and is supported only on (EC)  
476 transformations of  $\mathcal{U}$ .

477 **Definition 33. Strictly (EC) convolution operator**

478 A *strictly* edge-constrained (EC\*) convolution operator  $f_w$ , is an (EC) con-  
479 volution operator such that its supporting set  $\mathcal{N} \subset \mathcal{U}$ .

480 Let  $f_w$  be an (EC\*) convolutional operator. In the general case (i),  $w \in \mathcal{S}(V)$ ,  
481 so its support is  $\mathcal{M} = \varphi(\mathcal{N})$  such that  $\mathcal{N} \subseteq \mathcal{U}$ . In the abelian case (ii), we  
482 use instead  $w \in \mathcal{S}(\Gamma)$ , and thus its support is directly  $\mathcal{N}$ . Therefore, we can  
483 rewrite the expressions of the convolution operator as:

484 (i)  $\forall s \in \mathcal{S}(V), \forall u \in V, f_w(s)[u] \stackrel{(\varphi)}{=} \sum_{v \in \mathcal{K}_u} s[v] w[g_v^{-1}(u)]$

485 (ii)  $\forall s \in \mathcal{S}(V), f_w(s) \stackrel{(M)}{=} \sum_{g \in \mathcal{N}} w[g] g(s)$

486 *Remark.* Note that in the abelian case, we can see from (ii) that a definition  
 487 of a local patch would coincide with the supporting set, so that locality and  
 488 weight sharing is straightforward.

489 From these expressions, it is clear that  $\Gamma$  needs not to be fully determined  
 490 to calculate  $f_w(s)[u]$ . The case (ii) is the simplest as the only requirement  
 491 is a supporting set  $\mathcal{N}$  of (EC) invertible transformations. In the case (i), we  
 492 only need to determine  $\mathcal{K}_u$ .

493 **What exactly is the local patch  $\mathcal{K}_u$  ?**

494 **TODO: or not todo?**

495

## 2.4 From groups to groupoids

### 2.4.1 Motivation

One possible limitation coming from searching for Cayley subgraphs is that they are order-regular *i.e.* the in- and the out-degree  $d = |\mathcal{U}|$  of each vertex is the same. That is, for a general graph  $G$ , the size of the weight kernel  $w$  of an (EC\*) convolution operator  $f_w$  supported on  $\mathcal{U}$  is bounded by  $d$ , which in turn is bounded by twice the minimal degree of  $G$  (twice because  $G$  is undirected and  $\mathcal{U}$  can contain every inverse).

There are a lot of possible strategies to overcome this limitation. For example:

1. connecting each vertex with its  $k$ -hop neighbors, with  $k > 1$ ,
2. artificially creating new connections for less connected vertices,
3. allowing the supporting set  $\mathcal{N}$  to exceed  $\mathcal{U}$  *i.e.* dropping \* in (EC\*).

These strategies require to concede that the topological structure supported by  $G$  is not the best one to support an (EC\*) convolution on it, which breeds the following question:

- What can we relax in the previous (EC\*) construction in order to unbound the supporting set, and still preserve the equivariance characterization?

The latter constraint is a consequence that every vertex of the Cayley subgraph  $\vec{G}$  must be composable with every generator from  $\mathcal{U}$ . Therefore, the answer consists in considering groupoids (Brandt, 1927) instead of groups. Roughly speaking, a groupoid is almost a group except that its composition law needs not be defined everywhere. Weinstein, 1996, unveiled the benefits to base convolutions on groupoids instead of groups in order to exploit partial symmetries.

### 2.4.2 Definitions of some groupoid related notions



522 **Definition 34. Groupoid**

523 A *groupoid*  $\Upsilon$  is a set equipped with a partial composition law with domain  
524  $\mathcal{D} \subset \Upsilon \times \Upsilon$  that is

525 1. closed into  $\Upsilon$  i.e.  $\forall (g, h) \in \mathcal{D}, gh \in \Upsilon$

526 2. associative i.e.  $\forall f, g, h \in \Upsilon, \begin{cases} (f, g), (g, h) \in \mathcal{D} \Leftrightarrow (fg, h), (f, gh) \in \mathcal{D} \\ (f, g), (fg, h) \in \mathcal{D} \Leftrightarrow (g, h), (f, gh) \in \mathcal{D} \\ \text{when defined, } (fg)h = f(gh) \end{cases}$

527 3. invertible i.e.  $\forall g \in \Upsilon, \exists ! g^{-1} \in \Upsilon$  s.t.  $\begin{cases} (g, g^{-1}), (g^{-1}, g) \in \mathcal{D} \\ (g, h) \in \mathcal{D} \Rightarrow g^{-1}gh = h \\ (h, g) \in \mathcal{D} \Rightarrow hgg^{-1} = h \end{cases}$

529 Optionally, it can be *domain-symmetric* i.e.  $(g, h) \in \mathcal{D} \Leftrightarrow (h, g) \in \mathcal{D}$ , and  
530 *abelian* i.e. domain-symmetric with  $gh = hg$ .

531 *Remark.* Note that left and right inverses are necessarily equal (because  
532  $(gg^{-1})g = g(g^{-1}g)$ ). Also note we can define a right identity element  $e_g^r =$   
533  $g^{-1}g$ , and a left one  $e_g^l = gg^{-1}$ , but they are not necessarily equal and depend  
534 on  $g$ .

535 Most definitions related to groups can be adapted to groupoids. In particular,  
536 let's adapt a few notions.

537 **Definition 35. Groupoid partial action**

538 A partial *action* of a groupoid  $\Upsilon$  on a set  $V$ , is a function  $L$ , with domain  
539  $\mathcal{D}_L \subset \Upsilon \times V$  and valued in  $V$ , such that the map  $g \mapsto L_g$  is a groupoid  
540 homomorphism.

541 *Remark.* As usual, we will confound  $L_g$  and  $g$  when there is no possible  
 542 confusion, and we denote  $\mathcal{D}_{L_g} = \mathcal{D}_g = \{v \in V, (g, v) \in \mathcal{D}_L\}$ .

543 **Definition 36. Partial equivariant map**

544 A map  $\varphi$  from a groupoid  $\Upsilon$  partially acting on the destination set  $V$  is said  
 545 to be a *partial equivariant map* if

$$\forall g, h \in \Upsilon, \begin{cases} \varphi(h) \in \mathcal{D}_g \Leftrightarrow (g, h) \in \mathcal{D} \\ g(\varphi(h)) = \varphi(gh) \end{cases}$$

546 Also,  $\varphi$ -equivalence between a subgroupoid and a set is defined similarly with  
 547  $\varphi$  being a bijective *partial equivariant map* between them.

548 **Partial transformations groupoid**

549 We denote by  $\Psi(V)$  the set of partially defined transformations on  $v$ , and  
 550  $\Psi^*(V)$  the groupoid of its invertible ones (*i.e.* injective ones).

551 **2.4.3 Construction of the groupoid convolution**

552 The expression of the convolution we constructed in the previous section  
 553 cannot be applied as is. We first need to extend the algebraic objects we  
 554 work with. Extending a partial transformation  $g$  on the signal space  $\mathcal{S}(V)$   
 555 (and thus the convolutions) is a bit tricky, because only the signal entries  
 556 corresponding to  $\mathcal{D}_g$  are moved. A convenient way to do this is to consider  
 557 the groupoid closure obtained with the addition of an absorbing element

558 **Definition 37. Zero-closure**

559 The *zero-closure* of a groupoid  $\Upsilon$ , denoted  $\Upsilon^0$ , is the set  $\Upsilon \cup 0$ , such that the  
 560 groupoid axioms 1, 2 and 3, and the domain  $\mathcal{D}$  are left unchanged, and

561 4. the composition law is extended to  $\Upsilon^0 \times \Upsilon^0$  with  $\forall (g, h) \notin \mathcal{D}, gh = 0$

562 *Remark.* Note that this is coherent as the properties 2 and 3 are still partially  
 563 defined on the original domain  $\mathcal{D}$ .

564 Now, we will also extend every other algebraic object used in the expression  
 565 of the  $\varphi$ -convolution and the M-convolution, so that we can directly apply  
 566 our previous constructions.

567 **Lemma 38. Extension of  $\varphi$  on  $V^0$**

568 Let a partial equivariant map  $\varphi : \Upsilon \rightarrow V$ . It can be extended to a (total)  
 569 equivariant map  $\varphi : \Upsilon^0 \rightarrow V^0 = V \cup \varphi(0)$ , such that  $\varphi(0) \notin V$ , that we  
 570 denote  $0_V = \varphi(0)$ , and such that

$$\forall g \in \Upsilon^0, \forall v \in V^0, g(v) = \begin{cases} \varphi(gg_v) & \text{if } g_v \in \mathcal{D}_g \\ 0_V & \text{else} \end{cases}$$

571 *Proof.* We have  $\varphi(0) \notin V$  because  $\varphi$  is bijective. Additionally, we must have  
 572  $\forall (g, h) \notin \mathcal{D}, g(\varphi(h)) = \varphi(gh) = \varphi(0) = 0_V$ .  $\square$

573 *Remark.* Note that for notational conveniency, we may use the same symbol 0  
 574 for  $0_\Upsilon$ ,  $0_V$  and  $0_{\mathbb{R}}$ .

575 **Lemma 39. Extension of the subgroupoid actions on  $\mathcal{S}(V)$**

576 Let a subgroupoid  $\Upsilon \subset \Psi^*(V)$  such that  $\Upsilon \stackrel{\varphi}{\cong} V$ . Under the convention that  
 577  $s[0_V] = 0_{\mathbb{R}}$ , a transformation  $g \in \Upsilon^0$  can be extended to the signal space  
 578  $\mathcal{S}(V)$  as usual:

$$\forall s \in \mathcal{S}(V), \forall v \in V, g(s)[v] = \begin{cases} s[g^{-1}(v)] & \text{if } g \neq 0_\Upsilon \text{ and } v \in \mathcal{D}_{g^{-1}} \\ 0_{\mathbb{R}} & \text{else} \end{cases}$$

579 *Proof.* Straightforward.  $\square$

580 With these extensions, we can obtain the  $\varphi$ - and M-convolutions related to  $\Upsilon$   
 581 almost by substituting  $\Upsilon^0$  to  $\Gamma$  in Definition 18 and Definition 20.

582 **Definition 40. Groupoid convolution**

583 Let a subgroupoid  $\Upsilon \subset \Psi^*(V)$ , such that  $\Upsilon \stackrel{\varphi}{=} V$ . The  $\varphi$ - and M-convolutions,  
 584 based on  $\Upsilon$ , are defined on its zero-closure, with the same expression as if  $\Upsilon^0$   
 585 were a subgroup, and by extension of  $\varphi$  and of the groupoid actions *i.e.*

586 (i)  $\forall s, w \in \mathcal{S}(V), s *_{\varphi} w = \sum_{v \in V} s[v] g_v(w) = \sum_{g \in \Upsilon} s[\varphi(g)] g(w)$

587 (ii)  $\forall (w, s) \in \mathcal{S}(\Upsilon) \times \mathcal{S}(V), w *_{\text{M}} s = \sum_{g \in \Upsilon} w[g] g(s)$

588 where (ii) applies in the abelian case.

589 **Symmetrical expressions**

590 Note that, as  $\forall r, r[0] = 0$ , the groupoid convolutions can also be expressed  
 591 on the domain  $\mathcal{D}$  with a convenient symmetrical expression:

592 (i)  $\forall u \in V, (s *_{\varphi} w)[u] = \sum_{\substack{(g_a, g_b) \in \mathcal{D} \\ \text{s.t. } g_a g_b = g_u}} s[a] w[b]$

593 (ii)  $\forall u \in V, (w *_{\text{M}} s)[u] = \sum_{\substack{v \in \mathcal{D}_g \\ \text{s.t. } g(v) = u}} w[g] s[v]$

594 We obtain an equivariance characterization, similarly to Proposition 19 and  
 595 Corrolary 24.

596 **Proposition 41. Characterization by equivariance to  $\Upsilon$**

597 Let a subgroupoid  $\Upsilon \subset \Psi^*(V)$ , such that  $\Upsilon \stackrel{\varphi}{=} V$ . Then,

598 (i) right  $\varphi$ -convolution operators are equivariant to  $\Upsilon$ ,

599 (ii) if  $\Upsilon$  is abelian, left M-convolution operators are equivariant to  $\Upsilon$ .

600 Conversely,

601 (i) if  $\Upsilon$  is domain-symmetric, linear transformations of  $\mathcal{S}(V)$  that are  
 602 equivariant to  $\Upsilon$  are right  $\varphi$ -convolution operators,

- 603 (ii) if  $\Upsilon$  is abelian, linear transformations of  $\mathcal{S}(V)$  that are equivariant to  $\Upsilon$   
 604 are also left  $M$ -convolution operators.

605 *Proof.* (i) (a) Direct sense:

606 Using the symmetrical expressions, and the fact that  $\forall r, r[0] = 0$ ,  
 607 we have

$$\begin{aligned}
 (f_w \circ g(s))[u] &= \sum_{\substack{(g_a, g_b) \in \mathcal{D} \\ s.t. \ g_a g_b = g_u}} g(s)[a] w[b] \\
 &= \sum_{\substack{(g_a, g_b) \in \mathcal{D} \\ s.t. \ g_a g_b = g_u}} s[g^{-1}(a)] w[b] \\
 &= \sum_{\substack{(g_a, g_b) \in \mathcal{D} \\ s.t. \ (g, g_a) \in \mathcal{D} \\ s.t. \ g g_a g_b = g_u}} s[a] w[b] \\
 &= \sum_{\substack{(g_a, g_b) \in \mathcal{D} \\ s.t. \ (g, g_a) \in \mathcal{D} \\ s.t. \ g_a g_b = g^{-1} g_u = g_{\varphi(g^{-1} g_u)} = g_{g^{-1}(u)}}} s[a] w[b] \\
 &= f_w(s)[g^{-1}(u)] \\
 &= (g \circ f_w(s))[u]
 \end{aligned}$$

608 (b) Converse:

609 Let  $v \in V$ . Denote  $e_{g_v}^r = g_v^{-1} g_v$  the right identity element of  $g_v$ ,  
 610 and  $e_v^r = \varphi(e_{g_v}^r)$ . We have that

$$\begin{aligned}
 g_v(e_v^r) &= v \\
 \text{So, } \delta_v &= g_v(\delta_{e_v^r})
 \end{aligned}$$

611 Let  $f \in \mathcal{L}(\mathcal{S}(V))$  that is equivariant to  $\Upsilon$ , and  $s \in \mathcal{S}(V)$ . Thanks

612 to the previous remark we obtain that

$$\begin{aligned}
 f(s) &= \sum_{v \in V} s[v] f(\delta_v) \\
 &= \sum_{v \in V} s[v] f(g_v(\delta_{e_v^r})) \\
 &= \sum_{v \in V} s[v] g_v(f(\delta_{e_v^r})) \\
 &= \sum_{v \in V} s[v] g_v(w_v)
 \end{aligned} \tag{13}$$

613 where  $w_v = f(\delta_{e_v^r})$ . In order to finish the proof, we need to find  $w$   
 614 such that  $\forall v \in V, g_v(w) = g_v(w_v)$ .

615 Let's consider the equivalence relation  $\mathcal{R}$  defined on  $V \times V$  such  
 616 that:

$$\begin{aligned}
 a\mathcal{R}b &\Leftrightarrow w_a = w_b \\
 &\Leftrightarrow e_a^r = e_b^r \\
 &\Leftrightarrow g_a^{-1}g_a = g_b^{-1}g_b \\
 &\Leftrightarrow (g_b, g_a^{-1}) \in \mathcal{D} \\
 &\Leftrightarrow (g_a^{-1}, g_b) \in \mathcal{D}
 \end{aligned} \tag{14}$$

617 with (14) owing to the fact that  $\Upsilon$  is domain-symmetric.

618 Given  $x \in V$ , denote its equivalence class  $\mathcal{R}(x)$ . Under the hy-  
 619 pothesis of the axiom of choice (Zermelo, 1904) (if  $V$  is infinite),  
 620 define the set  $\aleph$  that contains exactly one representative per equiv-  
 621 alence class. Let  $w = \sum_{n \in \aleph} w_n$ . Then  $V$  is the disjoint union  
 622  $V = \cup_{n \in \aleph} \mathcal{R}(n)$  and (13) rewrites:

$$\forall u \in V, f(s)[u] = \sum_{n \in \aleph} \sum_{v \in \mathcal{R}(n)} s[v] g_v(w_n)[u]$$

$$\begin{aligned}
&= \sum_{n \in \mathbb{N}} \sum_{v \in \mathcal{R}(n)} s[v] w_n[g_v^{-1}(u)] \\
&= \sum_{n \in \mathbb{N}} \sum_{v \in \mathcal{R}(n)} s[v] w[g_v^{-1}(u)] \\
&= (s *_{\varphi} w)[u]
\end{aligned} \tag{15}$$

where (15) is obtained thanks to (14).

(ii) With symmetrical expressions, it is clear that the convolution is abelian, if and only if,  $\Upsilon$  is abelian. Then (i) concludes.

□

#### 2.4.4 Edge-constrained groupoid convolutions

Similarly to the construction of 2.3.1, we start by adapting the related notions.

##### Definition 42. Groupoid generating set

A set  $\mathcal{U}$ , equipped with a partial composition law of domain  $\mathcal{D}_0$ , is a *generating set* of a groupoid  $\Upsilon$ , if every object of  $\Upsilon$  can be expressed as a composition of objects of  $\mathcal{U}$  and their inverses. In this case, the domain  $\mathcal{D}$  of the partial composition law of  $\Upsilon$  can be deducted inductively from  $\mathcal{D}_0$  with the associativity and invertibility axioms of  $\Upsilon$  (and eventually also with the domain-symmetric axiom if  $\Upsilon$  is domain-symmetric).

A partial Cayley graph is defined similarly as the (total) Cayley graph.

##### Definition 43. Partial Cayley graph

Let a groupoid  $\Upsilon = \langle \mathcal{U}, \mathcal{D}_0 \rangle$ . The *partial Cayley graph* generated by  $\mathcal{U}$  and  $\mathcal{D}_0$ , is the digraph  $\vec{G} = \langle V, E \rangle$  such that  $V = \Upsilon$  and  $E$  is such that:

$$u \rightarrow v \Leftrightarrow \exists g \in \mathcal{U}, ga = b$$

641 We call *partial Cayley subgraph*, a subgraph that is isomorph to a partial  
642 Cayley graph.

643 *Remark.* The characterization by a partial Cayley subgraph, of graphs ad-  
644 mitting an (EC) convolution based on a groupoid, similar to Proposition 27  
645 also holds for the groupoid representation. However, it is somewhat trivial  
646 as every graph admits a partial Cayley subgraph (for example by labelling  
647 every edge in a neighbour with a transformation).

648 In the case of the groupoid representation, what is more interesting is that  
649 an (EC) groupoid convolution can be characterized by a generating set of  
650 the groupoid it is based on.

651 **Proposition 44. Characterization by generating set**

652 Let a graph  $G = \langle V, E \rangle$  such that it admits an (EC) convolution based on  
653 a groupoid  $\Upsilon$ . Then,  $G$  contains a sugraph  $\vec{G}$  that is isomorph to a partial  
654 Cayley subgraph  $\vec{G}_c$ , of generating set  $\mathcal{U} \subset \Psi_{\text{EC}}^*(G)$ , such that the underly-  
655 ing equivariant map  $\varphi$  of the (EC) convolution is also a graph isomorphism  
656 between  $\vec{G}_c$  and  $\vec{G}$ .

657 The proof is omitted because it would be highly similar with the one of  
658 Corrolary 28.

659 **Strictly edge-constrained groupoid convolution**

660 Thanks to the previous construction, in order to define a meaningful con-  
661 volution operator on a general graph  $G = \langle V, E \rangle$ , all we need is a set  $\mathcal{U}$   
662 of (EC) partial transformations *that need not to be defined on every vertex*,  
663 and a composition rule  $\mathcal{D}_0$ . From the set  $\mathcal{U}$ , we can choose a supporting set  
664  $\mathcal{N} \subset \mathcal{U}$  (from which we eventually need to derive the local patches  $\mathcal{K}_u$ ), and  
665 obtain the (EC\*) convolution operator  $f_w$ , without the need of fully deter-  
666 mining  $\Upsilon = \langle \mathcal{U} \rangle$ . We can choose between three levels of constraints, which  
667 are inherited from  $\mathcal{U}$  and  $\mathcal{D}_0$ :



- 668 1.  $\Upsilon$  is unconstrained, then  $f_w$  is an (EC\*)  $\varphi$ -convolution right-operator  
669 that is equivariant to  $\Upsilon$ , but the converse doesn't hold,
- 670 2.  $\Upsilon$  is domain-symmetric, then  $f_w$  is an (EC\*)  $\varphi$ -convolution right oper-  
671 ator and the equivariance characterization holds,
- 672 3.  $\Upsilon$  is abelian, then  $f_w$  is an (EC\*) M-convolution operator, the equiv-  
673 ariance characterization holds, and we don't need to compute the local  
674 patches  $\mathcal{K}_u$ .

675 In the next chapter, we will encounter examples of convolutions defined on  
676 graphs, that can be expressed under group representations or under groupoid  
677 representations.



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