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# Chapter 2

## Convolutions on graph domains

### Introduction

Defining a convolution of signals over graph domains is a challenging problem. If the graph is not a grid graph, there exists no natural extension of the euclidean convolution.

In Section 2.1, we analyze the reasons why the euclidean convolution operator is useful in deep learning, and give a characterization. Then we will search for domains onto which a convolution with these properties can be naturally obtained.

This will lead us to put our interest on representation theory and convolutions defined on groups in Section 2.2. As the euclidean convolution is just a particular case of the group convolution, it makes perfect sense to steer our construction in this direction. Hence, we will aim at transferring its representation on the vertex domain.

Then, in Section 2.3, we will introduce the role of the edge set and see how it should influence it. This will provide us with some particular classes of graphs for which we will obtain a natural construction with the wanted characteristics that we exposed in the first place.

Finally, we will relax some aspect of the construction to adapt it to general graphs in Section 2.4. The obtained construction is a set of general expressions that describes convolutions on graph domains and preserves some key properties.

We summarize our constructions in a conclusive Section 2.5.

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## 69 2.1 Analysis of the classical convolution

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

### 74 2.1.1 Properties of the convolution

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

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

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

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

#### 80 **Definition 2. Convolution operator**

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

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

#### 87 **Linearity**

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

90 **Locality and weight sharing**

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

94 **Commutativity**

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

97 **Equivariance to translations**

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

100 **2.1.2 Characterization on grid graphs**

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

102 **Definition 3. Transformation**

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

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

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

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

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

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

112 Then, we have:

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

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

113

□

114 We also recall the formalism of translations.

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

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

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

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

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

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

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

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

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

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

125

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

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

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

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

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

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

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



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

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

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

133

□

### 134 2.1.3 Usefulness of convolutions in deep learning

#### 135 Equivariance property of CNNs

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

#### 140 Lossless superiority of CNNs over MLPs

141 The converse result, as a consequence of Proposition 6, is never mentioned  
 142 in deep learning literature. However it is also a strong one. For example,  
 143 let's consider a linear function that is equivariant to translations. Thanks  
 144 to the converse result, we know that this function is a convolution operator  
 145 parameterized by a weight vector  $w$ ,  $f_w : \cdot * w$ . If the domain is compactly  
 146 supported, as in the case of images, we can break down the information of  $w$   
 147 in a finite number  $n_q$  of kernels  $w_q$  with small compact supports of same size  
 148 (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  
 149 convolution operators  $f_{w_q}$  are all in the search space of  $2 \times 2$  convolutional  
 150 layers. In other words, every translational equivariant linear function can

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

155 **Methodology for extending to general graphs**

156 Hence, in our construction, we will try to preserve the characterization from  
157 Proposition 6 as it is mostly the reason why they are successful in deep  
158 learning. Note that the reduction of parameters compared to a dense layer  
159 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.

186 **Definition 8. Group convolution I**

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

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

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

190 **2.2.1 Steered construction from groups**

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

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

197 Then, the linear morphism  $\tilde{\varphi}$  from  $\mathcal{S}(\Gamma)$  to  $\mathcal{S}(V)$  defined on the Dirac bases  
 198 by  $\tilde{\varphi}(\delta_g) = \delta_{\varphi(g)}$  is a linear isomorphism. Hence,  $\mathcal{S}(V)$  would inherit the same  
 199 inherent structural properties as  $\mathcal{S}(\Gamma)$ . For the sake of notational simplicity,  
 200 we will use the same symbol  $\varphi$  for both  $\varphi$  and  $\tilde{\varphi}$  (as done between  $f$  and  
 201  $L_f$ ). 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

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

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

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

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

207

□

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

210 **Definition 10. Group convolution II**

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

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

213

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

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

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

216

217 *Proof.* Using Lemma 9,

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

218

□

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

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

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

if  $\varphi \in \text{ISO}(\Gamma, V)$ ,

$$\exists w \in \mathcal{S}(V), f = . *_{\text{II}} w \Rightarrow \forall v \in V, f \circ \varphi(g_v) = \varphi(g_v) \circ f$$

223

*Proof.*

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

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

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

225 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:

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

226

□

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

230 In fact, both group convolutions are the same as the latter one borrows the  
 231 algebraic structure of the first one. Thus we only obtain equivariance to  $\varphi(\Gamma)$   
 232 when  $\varphi$  also transfer the group structure from  $\Gamma$  to  $V$ , and the converse does  
 233 not hold. To obtain equivariance to  $\Gamma$  (and its converse), we will drop the  
 234 direct homomorphism condition, and instead we will take into account the  
 235 fact that it contains invertible transformations of  $V$ .

## 2.2.2 Construction under group actions

### Definition 13. Group action

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

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

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

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

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

### Definition 14. Equivariant map

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

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

251

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

253



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

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

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

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

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

261 **Definition 16. Group convolution III**

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

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

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

264

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

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

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

*Proof.*

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

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

270 Hence, the direct sense is obtained by applying (P).

271 For the converse, given  $u, v \in V$ , we first realize (5) for  $s_1 := \delta_v$ , obtaining  
 272  $s_2[\varphi(g_v^{-1}g_u)] = s_2[g_v^{-1}(u)]$ , which we then realize for a real signal  $s_2$  having no  
 273 two equal entries, obtaining  $\varphi(g_v^{-1}g_u) = g_v^{-1}(u)$ . From the latter we finally  
 274 obtain (P) with the one-to-one correspondence  $g_{v'} := g_v^{-1}$ .  $\square$

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

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

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

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

279

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

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

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

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

286

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

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

290 because of (P).

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

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

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

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

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

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

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

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

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

292 2. From right to left:

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

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

295 Thanks to (P), we have that:

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

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

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

296 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  $\varphi$ .

312 **Definition 20. Mixed domain convolution**

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

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

316

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

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

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

321 
$$r *_{\text{M}} s = \varphi(r) *_{\text{III}} s$$

322

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

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

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

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

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

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

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

329

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

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

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

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

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

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

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

349 by particularizing on well chosen signals):

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

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

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

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

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

350

□

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

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

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

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

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

358

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

361 Depending on the applications, we will build upon either  $*_{\varphi}$  or  $*_{\mathbf{M}}$  when the  
 362 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.

### 2.3.1 Edge-constrained convolutions

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

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

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.



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

386 This leads us to consider Cayley graphs (Cayley, 1878).

387 **Definition 26. Cayley graph**

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

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

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

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

394 **Convolution on Cayley graphs**

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

398 More precisely, we obtain the following characterization:

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

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

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

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

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

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

408 2. From right to left:

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

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

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

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

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

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

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

422 and

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

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

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

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

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

428 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  
429 map  $\varphi$  as:

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

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

431 Additionally, we have:

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

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

435

□

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

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

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

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

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

### 441 2.3.2 Intrinsic properties

442 • Obviously the constructed convolutions are linear. But do they also  
443 preserve the locality and weight sharing properties ?

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

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

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

### 449 **Proposition 29. Locality**

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

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

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

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

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

463 **Definition 30. Supporting set**

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

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

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

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

471 *Remark.* In other terms,  $\mathcal{K}_{\text{Id}} = \varphi(\mathcal{N}^{-1})$  is the *initial local patch*, which is  
 472 composed of all vertices that are connected in direction to  $\varphi(\text{Id})$ ; and  $\mathcal{K}_u$  is  
 473 obtained by moving  $\mathcal{K}_{\text{Id}}$  on the Cayley subgraph via the edges corresponding  
 474 to the decomposition of  $g_u$  on the generating set  $\mathcal{U}$ .

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

477 **Proposition 32. Weight sharing**

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

479 *Proof.* By using (P),

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

480

□

### 481 2.3.3 Stricly edge-constrained convolutions

482 We make the distinction between general (EC) convolution operators and  
 483 those for which the weight kernel  $w$  is smaller and is supported only on (EC)  
 484 transformations of  $\mathcal{U}$ .

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

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

488 *Remark.* (EC\*) convolution operators are simpler to obtain as we can con-  
 489 struct them just with  $\mathcal{U} \subset \Phi_{\text{EC}}^*(G)$  without composing the transformations.

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

$$494 \quad \text{(i)} \quad \forall s \in \mathcal{S}(V), \forall u \in V, f_w(s)[u] \stackrel{(\varphi)}{=} \sum_{v \in \mathcal{K}_u} s[v] w[g_v^{-1}(u)]$$

$$495 \quad \text{(ii)} \quad \forall s \in \mathcal{S}(V), f_w(s) \stackrel{(\text{M})}{=} \sum_{g \in \mathcal{N}} w[g] g(s)$$

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

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

## 503 2.4 From groups to groupoids

### 504 2.4.1 Motivation

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

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

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

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

- 518 • What can we relax in the previous (EC\*) construction in order to un-  
 519 bound the supporting set, and still preserve the equivariance charac-  
 520 terization?

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



## 2.4.2 Definition of notions related to groupoids

### Definition 34. Groupoid

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

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

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

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

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

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

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

### Definition 35. Groupoid partial action

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

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

550 **Definition 36. Partial equivariant map**

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

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

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

555 **Definition 37. Partial transformations groupoid**

556 The *partial transformations groupoid*  $\Psi^*(V)$ , is the set of invertible par-  
 557 tial transformations, equipped with the functional composition law with do-  
 558 main  $\mathcal{D}$  such that

$$\begin{cases} \mathcal{D}_{gh} = h(\mathcal{D}_h) \cap \mathcal{D}_g \\ (g, h) \in \mathcal{D} \Leftrightarrow \mathcal{D}_{gh} \neq \emptyset \end{cases}$$

559 *Remark.* Note that a subgroupoid  $\Upsilon \subset \Psi^*(V)$  is domain-symmetric when  
 560  $\exists v \in V, g(v) \in \mathcal{D}_h \Leftrightarrow \exists u \in V, h(u) \in \mathcal{D}_g$

561 **2.4.3 Construction of partial convolutions**

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

**Definition 38. Zero-closure**

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

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

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

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

**Lemma 39. Extension of  $\varphi$  on  $V^0$** 

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

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

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

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

Similarly to  $\Phi^*(V)$ ,  $\Psi^*(V)$  can also move signals of  $\mathcal{S}(V)$ .

**Lemma 40. Extension of injective partial transformations to  $\mathcal{S}(V)$** 

Let  $g \in \Psi^*(V)$ . Its extension is done in two steps:

1.  $g$  is extended to  $V^0 = V \cup \{0_V\}$  as  $g(v) = 0_V \Leftrightarrow v \notin \mathcal{D}_g$ .

589 2. Under the convention  $\forall s \in \mathcal{S}(V), s[0_V] = 0_{\mathbb{R}}$ ,  $g$  is extended via linear  
 590 extension to  $\mathcal{S}(V)$ , and we have

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

591 *Proof.* Straightforward. □

592 With these extensions, we can obtain the partial  $\varphi$ - and M-convolutions re-  
 593 lated to  $\Upsilon$  almost by substituting  $\Upsilon^0$  to  $\Gamma$  in Definition 18 and Definition 20.

594 **Definition 41. Partial convolution**

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

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

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

601 **Symmetrical expressions**

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

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

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

606 We obtain an equivariance characterization similar to Proposition 19 and  
 607 Corrolary 24.

**Proposition 42. Characterization by equivariance to  $\Upsilon$** 

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

1. Then,

- (i) partial  $\varphi$ -convolution right-operators are equivariant to  $\Upsilon$ ,
- (ii) if  $\Upsilon$  is abelian, partial M-convolution left-operators are equiv to  $\Upsilon$ .

2. Conversely,

- (i) if  $\Upsilon$  is domain-symmetric, linear transformations of  $\mathcal{S}(V)$  that are equivariant to  $\Upsilon$  are partial  $\varphi$ -convolution right-operators,
- (ii) if  $\Upsilon$  is abelian, they are also partial M-convolution left-operators.

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

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

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

620 (b) Converse:

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

$$g_v(e_v^r) = v$$

$$\text{So, } \delta_v = g_v(\delta_{e_v^r})$$

623 Let  $f \in \mathcal{L}(\mathcal{S}(V))$  that is equivariant to  $\Upsilon$ , and  $s \in \mathcal{S}(V)$ . Thanks  
 624 to the previous remark we obtain that

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

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

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

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

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

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

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

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.

□

### Inclusion of (EC)

Similarly to the construction in Section 2.3, partial convolutions can define (EC) and (EC\*) counterparts with a characterization of admissibility by groupoid Cayley subgraph isomorphism, and similar intrinsic properties.

### Limitation of partial convolutions

However, because of the groupoid associativity, if  $g \in \Psi_{\text{EC}}^*(G)$ , then, any  $v \in V$  s.t.  $g(u) = v$  would be constrained to allow to be acted by every  $h$  s.t.  $(h, g) \in \mathcal{D}$ , which fails at unbounding the supporting set of a partial (EC\*) convolutions.

#### 2.4.4 Construction of path convolutions

To answer the limitation of partial convolutions, given  $g \in \langle \mathcal{U} \rangle$  where  $\mathcal{U} \subset \Psi_{\text{EC}}^*(G)$ , the idea is to proceed with a foliation of  $g$  into pieces, each corresponding to an edge  $e \in E$ , and together generating another groupoid with a different associativity law, as follows.

##### Definition 43. Path groupoid

Let  $\mathcal{U} \subset \Psi_{\text{EC}}^*(G)$ . The *path groupoid* generated from  $\mathcal{U}$ , denoted  $\mathcal{U} \ltimes G$ , with composition rule  $\mathcal{D}_{\ltimes}$ , is the groupoid obtained inductively with:

1.  $\mathcal{U} \ltimes_1 G = \{(g, v) \in \mathcal{U} \times V, v \in \mathcal{D}_g\} \subset \mathcal{U} \ltimes G$
2.  $((g_n, v_n) \cdots (g_1, v_1), (h_m, u_m) \cdots (h_1, u_1)) \in \mathcal{D}_{\ltimes} \Leftrightarrow h_m(u_m) = v_1$
3.  $((g_n, v_n) \cdots (g_1, v_1))^{-1} = (g_1^{-1}, g_1(v_1)) \cdots (g_n^{-1}, g_n(v_n))$

Call path its objects. Given a length  $l \in \mathbb{N}$ , denote  $\mathcal{U} \ltimes_l G$  the subset composed of the paths that are the composition of exactly  $l$  paths of  $\mathcal{U} \ltimes_1 G$ .

*Remark.* This groupoid construction is inspired from the field of operator algebra where partial action groupoids have been extensively studied, *e.g.* Nica, 1994; Exel, 1998; Li, 2016.

Such groupoids usually come equipped with source and target maps. We also define the path map.

##### Definition 44. Source, target and path maps

Let a path groupoid  $\mathcal{U} \ltimes G$ . We define on it the *source map*  $\alpha$  the *target map*  $\beta$  and the *path map*  $\gamma$  as:

$$\begin{cases} \alpha : (g_n, v_n) \cdots (g_1, v_1) \mapsto v_1 \in V \\ \beta : (g_n, v_n) \cdots (g_1, v_1) \mapsto g_n(v_n) \in V \\ \gamma : (g_n, v_n) \cdots (g_1, v_1) \mapsto g_n g_{n-1} \cdots g_1 \in \Psi^*(V^0) \end{cases}$$



669 *Remark.* Note that the path groupoid can also be obtained by derivation of  
 670 the partial transformation groupoid (*i.e.*  $p \in \mathcal{U} \ltimes G$  can be seen as a derivative  
 671 of  $\gamma(p)$  *w.r.t.*  $\alpha(p)$ ), and can thus be seen as the local structure of it.

672 **Lemma 45.**

673 Note the following properties:

- 674 1.  $(p, q) \in \mathcal{D}_\ltimes \Leftrightarrow \alpha(p) = \beta(q)$
- 675 2.  $\alpha(p) = \beta(p^{-1})$
- 676 3.  $e_p^l = pp^{-1} = (\text{Id}, \beta(p))$  and  $e_p^r = p^{-1}p = (\text{Id}, \alpha(p))$
- 677 4.  $\gamma$  is a groupoid partial action. We will denote  $\gamma_p$  instead of  $\gamma(p)$ .

678 *Remark.* Note that this time we won't use the notation  $p(v)$  for  $\gamma_p(v)$  in order  
 679 to better differentiate between the composition laws in  $\langle \mathcal{U} \rangle$  and  $\mathcal{U} \ltimes G$ .

680 One of the key object of our contruction is the use of  $\varphi$ -equivalence in order  
 681 to transform a sum over a group(oid) of (partial) transformations, into a sum  
 682 over the vertex set. With the current notion of path groupoid, searching for  
 683 something similar amounts to searching for a graph traversal.

684 **Definition 46. Traversal set**

685 Let a graph  $G = \langle V, E \rangle$  that is connected. A *traversal set* is a pair  $(\mathcal{U}, \mathcal{T})$  of  
 686 (EC) partial transformations subsets  $\subset \Psi_{\text{EC}}^*(G)$ , such that

- 687 1. An edge can only correspond to a unique  $g \in \mathcal{U}$ ,  
 688 *i.e.*  $\forall g, h \in \mathcal{U} : \exists v \in V, g(v) = h(v) \Rightarrow g = h$
- 689 2. The (EC) partial transformations of  $\mathcal{T}$  are restrictions of those of  $\mathcal{U}$ ,  
 690 *i.e.*  $\forall g \in \mathcal{U}, \exists! h \in \mathcal{T}, \begin{cases} \mathcal{D}_h \subset \mathcal{D}_g \\ \forall v \in \mathcal{D}_h, h(v) = g(v) \end{cases}$   
 691 (equivalently,  $\mathcal{T} \ltimes G$  is a path subgroupoid of  $\mathcal{U} \ltimes G$  *s.t.*  $|\mathcal{T}| = |\mathcal{U}|$ )
- 692 3. The subgraph  $G_{\mathcal{T}} = \langle V, \mathcal{T} \ltimes_1 G \rangle$  is a spanning tree of  $G$ .

693 We denote  $(\mathcal{U}, \mathcal{T}) \in \text{trav}(G)$ , and denote by  $r$  the root of  $G_{\mathcal{T}}$ .

694 *Remark.* The assumption that the graph  $G$  is connected doesn't lose gener-  
 695 ality as the construction can be replicated to each connected component in  
 696 the general case.

697 A traversal set  $(\mathcal{U}, \mathcal{T})$  defines a  $\varphi$ -equivalence between the  $\alpha$ -fiber of the  
 698 root  $r$  and the vertex set  $V$  as follows.

699 **Lemma 47. Path  $\varphi$ -Equivalence**

700 Let  $(\mathcal{U}, \mathcal{T}) \in \text{trav}(G)$ . Given  $v \in V$ , there exists a unique  $p_v \in \mathcal{T} \ltimes G$  such  
 701 that  $\alpha(p_v) = r$  and  $\beta(p_v) = v$ . Define  $\varphi : p_v \mapsto v$ . Then  $\varphi : \alpha_{\mathcal{T} \ltimes G}^{-1}\{r\} \rightarrow V$  is  
 702 a bijective partial equivariant map.

703 *Proof.* Bijectivity is a consequence of the spanning tree structure of  $\mathcal{T}$ . Equiv-  
 704 ariance because  $\gamma_{p_v}(u) = \gamma_{p_v} \gamma_{p_u}(r) = \gamma_{p_v p_u}(r) = \varphi(p_v p_u)$ .  $\square$

705 We can now define the convolution that is based on a path groupoid.

706 **Definition 48. Path convolution**

707 Let  $(\mathcal{U}, \mathcal{T}) \in \text{trav}(G)$ . The *path convolution* is the partial convolution based  
 708 on the path subgroupoid  $\mathcal{T} \ltimes G$ , which uses the groupoid partial action  
 709  $\gamma := \gamma^{\mathcal{U} \ltimes G}$  of the embedding groupoid  $\mathcal{U} \ltimes G$ .

710 (i) In what follows are the three expressions of the path  $\varphi$ -convolution for  
 711 signals  $s_1, s_2 \in \mathcal{S}(V)$ , and  $u \in V$ :

$$\begin{aligned}
 (s *_{\varphi} w) &= \sum_{v \in V} s[v] \gamma_{p_v}(w) \\
 &= \sum_{\substack{p \in \mathcal{T} \ltimes G \\ \text{s.t. } \alpha(p)=r}} s[\varphi(p)] \gamma_p(w) \\
 (s *_{\varphi} w)[u] &= \sum_{\substack{(a,b) \in V \\ \text{s.t. } \gamma_{p_a}(b)=u}} s[a] w[b]
 \end{aligned}$$

712 (ii) The mixed formulations with  $w \in \mathcal{S}(\mathcal{T} \ltimes G)$  are:

$$\begin{aligned} (w *_{\mathcal{M}} s) &= \sum_{\substack{p \in \mathcal{T} \ltimes G \\ \text{s.t. } \alpha(p)=r}} w[p] \gamma_p(s) \\ (w *_{\mathcal{M}} s)[u] &= \sum_{\substack{(p,v) \in \mathcal{T} \ltimes G \times V \\ \text{s.t. } \alpha(p)=r \\ \text{s.t. } \gamma_p(v)=u}} w[p] s[v] \end{aligned}$$

713 *Remark.* The role of  $\mathcal{T}$  is to provide a  $\varphi$ -equivalence. The role of  $\mathcal{U}$  is to  
 714 extend every partial transformation  $\gamma_g^{\mathcal{T} \ltimes G}$  to the domain of its unrestricted  
 715 counterpart  $\gamma_g^{\mathcal{U} \ltimes G}$ .

716 Proposition 42 also holds for path groupoids, except that the domain-symmetric  
 717 condition of 2.(i) is not needed.

718 **Proposition 49. Characterization by equivariance to  $\mathcal{U} \ltimes G$ 's action**

719 Let  $(\mathcal{U}, \mathcal{T}) \in \text{trav}(G)$ .

- 720 (i) The class of linear transformations of  $\mathcal{S}(V)$  that are equivariant to the  
 721 path actions of  $\mathcal{U} \ltimes G$  is exactly the path  $\varphi$ -convolution right-operators;
- 722 (ii) in the abelian case, they are also exactly the M-convolution left-operators.

723 *Proof.* Instead of the domain-symmetric condition that was used in the proof  
 724 of the converse of Proposition 42 (2.(i)), we use the fact that any vertex can be  
 725 reached with an action from the root of the spanning tree of the traversal set.  
 726 Indeed, given  $v \in V$ , as we have  $\gamma_{p_v}(r) = v$ , then  $\gamma_{p_v}(\delta_r) = \delta_v$ . Therefore, by  
 727 developping a linear transformation  $f(s)$  on the dirac family, and commuting  
 728  $f$  with  $\gamma_{p_v}$ , we obtain that  $f(s) = s *_{\varphi} w$ , where  $w = f(\delta_r)$ . The rest of the  
 729 proof is similar to that of Proposition 42.  $\square$

730 *Remark.* Note that  $\mathcal{U} \ltimes V$ 's action is almost the same as the groupoid partial  
 731 action of  $\Upsilon = \langle \mathcal{U} \rangle$  (only "almost" because not all combinations of partial  
 732 transformations might exist in the paths). However  $\mathcal{U} \ltimes V$  associativity law  
 733 doesn't have the limitation of  $\Upsilon$ 's.

#### 734 **(EC\*) Path convolution operators**

735 The counterparts of strictly edge-constrained (EC\*) convolution operators  
 736 for path convolutions, are indeed path convolution operators obtained with  
 737 supporting set  $\mathcal{N} \subset \mathcal{T} \ltimes_1 G$  which any graph can admit. As shown by this  
 738 section, to construct one, all we need is a traversal set of partial transforma-  
 739 tions  $(\mathcal{U}, \mathcal{T})$ .

## 2.5 Conclusion

In this chapter, we constructed the convolution on graph domains.

1. We first saw that classical convolutions are in fact the class of linear transformations of the signal space that are equivariant to translations. For signals defined on graph domains, there is no natural definition of translations.
2. Therefore, we adopted a more abstract standpoint and considered in the first place any kind of transformation of the vertex set  $V$ . Hence, given a subgroup of transformation  $\Gamma$ , we constructed the class of linear transformations of the signal space that are equivariant to it. This provided us with an expression of a convolution based on this subgroup, and a bijective equivariant map between  $\Gamma$  and  $V$ , in order to transport a sum over  $\Gamma$  into a sum over  $V$ . We also proposed a simpler expression in the abelian case.
3. Then, we introduced the role of the edge set  $E$ , and we constrained  $\Gamma$  by it. This allows us to obtain a characterization of admissibility of convolutions by Cayley subgraph isomorphism, and to analyze intrinsic properties of the constructed convolution operator, namely locality and weight sharing. We also discussed operators with a smaller kernel, in particular those that are strictly edge-constrained (EC\*), as they are simpler to construct.
4. Finally, we overcame the limitation that some graphs only have trivials or low order Cayley subgraphs. In this case, we rebased our construction on groupoids of partial transformations  $\Upsilon$  as a first iteration, but this one didn't overcome fully the above-mentioned limitation. As a last iteration, we broke down the previous construction into elementary partial actions onto the edges, recomposed into path groupoids  $\mathcal{U} \ltimes G$ .

767 Similarly, equivariance characterization and intrinsic properties hold,  
 768 and the simpler (EC\*) construction is also possible.

769 **Summary of practical (EC\*) convolution operators**

770 3. For graphs that are quite regular, in the sense that they contain an  
 771 above-low-order Cayley subgraph (order  $k \geq 4$ ), we saw in Section 2.3.3  
 772 that all we need to construct an (EC\*) convolution operator is a gen-  
 773 erating set  $\mathcal{U}$  of transformations, without the need of composing its el-  
 774 ements, and optionally (in the non-abelian case) to move a local patch  
 775  $\mathcal{K}_{\text{Id}}$  over the graph domain.

776 4. For a general graph, we saw in Section 2.4.4 that all we need to con-  
 777 struct an (EC\*) path convolution operator is a traversal set  $(\mathcal{U}, \mathcal{T})$  of  
 778 partial transformations, without the need to compose the paths.

779 In the next chapter, we will encounter examples of (EC) and (EC\*) con-  
 780 volution operators defined on graphs, that can be expressed under group  
 781 representations or under path groupoid representations.

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