# Cliques in Association Schemes

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#### Abstract

This report is an exploration of some of the topics within *Delsarte theory*, which uses ideas from graph theory, algebra, and optimization to address questions in coding theory in particular, and combinatorics more broadly. The centre of study is the *association scheme*, which provides a setting in which to view various objects, especially *distance-regular graphs*. This perspective enables the computation of various parameters of interest, including the eigenvalues of graphs and upper bounds on codes, cliques, and independent sets.

This report aims to be mostly self-contained, though background knowledge of basic linear algebra and group theory is required. The important aspects of this theory is reviewed in the appendix.

# Contents

1	Inti	roduction	3		
	1.1	Coding Theory	3		
	1.2	Hamming Graphs	3		
	1.3	Distance-Regular Graphs	3		
2	Ass	ociation Schemes	4		
	2.1	Association Schemes	4		
		2.1.1 The Bose-Mesner Algebra	7		
		2.1.2 Duality	7		
	2.2	P-Polynomial Schemes	7		
		2.2.1 Q-Polynomial Schemes	7		
	2.3	Automorphisms and Cayley Graphs	7		
	2.4	Partitions and Translation Schemes	10		
	2.5	The Eigenvalues of the Hamming Scheme	14		
3	Delsarte's Linear Programming Bound				
	3.1	Linear Programming	17		
		3.1.1 Duality	18		
	3.2	The LP Bound	19		
	3.3	The Ratio Bound	21		
	3.4	The Clique-Coclique Bound	21		
4	Schrijver's SDP Bound				
	4.1	The Terwilliger Algebra of the Hamming Scheme	22		
	4.2	Semi-Definite Programming	22		
5	Cor	nputation	23		
6	Dis	cussion	24		
	6.1	Conclusion	24		

	6.2	Other Applications	24		
A	Linear Algebra				
	A.1	The Spectral Theorem	25		
	A.2	Adjacency Matrices	25		
	A.3	Positive Semi-Definite Matrices	25		
В	Gro	oup Theory	26		
	B.1	Group Actions	26		
	B.2	Character Theory	28		
	B.3	The Structure of Finite Abelian Groups	30		
$\mathbf{C}$		ation	32		
	C.1	Godsil	32		
	C.2	Delsarte	32		
	C.3	Schrijver	32		
Bi	Bibliography				

# 1. Introduction

# 1.1 Coding Theory

I'm also not sure if I should include a section like this, but seeing as coding theory was the original motivation behind Delsarte's LP bound, and it remains (presumably?) a strong motivator for this theory, I figured it might be interesting to mention this as an application.

- Basics of Coding Theory
- Linear Graphs
- Finite Vector Spaces

# 1.2 Hamming Graphs

# 1.3 Distance-Regular Graphs

Define M-cliques.

I'm not sure if I should maybe merge this section with P-polynomial section?

- Definition
- Basic parameters
- Examples?

# 2. Association Schemes

### 2.1 Association Schemes

- Definition(s)
- Examples?
- Basic parameters

**Definition 2.1.1** (Commutative Association Scheme – Combinatorial [1, Section 2.1]) Let  $D = \{0, 1, ..., d\}$  for some  $d \ge 1$ . A COMMUTATIVE ASSOCIATION SCHEME  $\mathcal{A}$  is a set X, called the VERTEX SET, together with a set of relations  $\{R_i\}_{i=0}^d$  satisfying the following axioms:

- 1. The set of relations  $\{R_i\}_{i=0}^d$  partitions  $X \times X$ ;
- 2.  $R_0$  is the diagonal relation  $\{(x,x) \mid x \in X\}$ ;
- 3. For each  $i \in D$ , there is an  $i' \in D$  such that  $R_{i'}$  is the opposite relation  $\{(y, x) \mid (x, y) \in R_i\}$  of  $R_i$ ;
- 4. For every triple  $i, j, k \in D$ , there exists a constant  $p_{i,j}^k$  such that for all  $(x,y) \in R_k$ , there are exactly  $p_{i,j}^k$  vertices z such that  $(x,z) \in R_i$  and  $(z,y) \in R_j$ ; furthermore,  $p_{i,j}^k = p_{j,i}^k$ .

As used above, the elements of X are called Vertices, and vertices  $(x,y) \in R_i$  are called  $i^{\text{TH}}$  ASSOCIATES.

Each relation  $R_i$  is called a CLASS of  $\mathcal{A}$ , which has DIAMETER d (this will be explained in connection with distance-regular graphs in the next section (2.2)). Sometimes,  $\mathcal{A}$  is said to be a d-class association scheme (as the diagonal relation is discounted).

To every relation  $R \subseteq X \times X$  there exists a  $X \times X$  01 matrix A, where

$$A_{xy} = \begin{cases} 1 & \text{if } (x,y) \in R \\ 0 & \text{if } (x,y) \notin R \end{cases}; \tag{2.1}$$

A is then called the ADJACENCY MATRIX of R. This is a bijective correspondence between relations on  $X \times X$ , and  $X \times X$  matrices with each entry either 0 or 1. (Note that binary relations are precisely directed graphs without parallel edges, and this usage of the term adjacency matrix agrees with its usage in graph theory.)

**Definition 2.1.2** (Commutative Association Scheme – Algebraic [3, Chapter 12]) Let  $\circ$  denote the SCHUR product of matrices of the same shape:

$$(A \circ B)_{xy} = A_{xy}B_{xy} . (2.2)$$

(This is also called the Hadamard, or entrywise product.) If each relation  $R_i$  has adjacency matrix  $A_i$ , the above axioms can be reformulated as follows:

- 1.  $\sum_{i=0}^{d} A_i = J$ , the all-ones matrix, and  $A_i \circ A_j = \delta_{i,j} A_i$ ;
- 2.  $A_0 = I$ , the identity matrix;
- 3. For every  $i \in D$  there is an  $i' \in D$  such that  $A_i^T = A_{i'}$ ;
- 4. For every i, j,

$$A_i A_j = A_j A_i = \sum_{k=0}^{d} p_{i,j}^k A_k$$
.

The first three equivalences are straightforward translations. To see the last, observe that

$$(A_i A_j)_{xy} = \sum_{z \in X} (A_i)_{xz} (A_j)_{zy}$$

which counts the number of vertices  $z \in X$  such that

$$\begin{cases} (A_i)xz = 1 & \iff (x, z) \in R_i \\ (A_j)zy = 1 & \iff (z, y) \in R_j \end{cases}.$$

Then,  $(A_iA_j)_{xy} = p_{ij}^k$  exactly when  $(A_k)_{xy} = 1 \iff (x,y) \in R_k$ .

The requirement in (Combinatorial 4) that  $p_{i,j}^k = p_{j,i}^k$  corresponds to the requirement that  $A_i A_j = A_j A_i$  (Algebraic 4), which is why such association schemes are called *commutative*.

If the requirement of *commutativity* is dropped, then every finite group G is an association scheme in the following way. Cayley's theorem says that every group is isomorphic to the group of permutations on G given by left multiplication (the same construction will work

if right multiplication is used everywhere instead of left). By identifying each element of G with the  $G \times G$  permutation matrix, we obtain an association scheme with G as the vertex set, and permutation matrices as the classes. Since the identity element of G will map to the identity matrix, the product of two such permutations is another such permutation, and the transpose of a permutation matrix is its inverse, axioms (2, 3, 4) are satisfied. Since for every  $g, h \in G$ , only the element  $hg^{-1}$  maps g to h, exactly one of the permutation matrices will have a 1 in the (g, h)-entry; all the others will be 0. This association scheme will be commutative if and only if the group G is.

Not every association scheme (commutative or not) arises is this way, but many schemes of interest are closely related to a particular group. Nevertheless, the connection between the combinatorial definition of association schemes (2.1.1) and the algebraic one (2.1.2) provides a rich connection between the two subjects (2.1.1).

A particularly important class of association scheme, called SYMMETRIC, is one in which every relation is symmetric (i.e. i = i' in Combinatorial 3), or equivalently, each adjacency matrix is symmetric ( $A_i^T = A_i$  in Algebraic 3). In this case, the relations  $R_i$  form graphs  $\Gamma_i$  with vertex set X, and edge set given by

$$x \sim y \iff (x, y) \in R_i \iff (y, x) \in R_i$$
.

The most important class of symmetric association scheme (for the purposes of this report) arise as the distance graphs of a distance-regular graph (2.2). From this setting, we can generalize the following definition. (TODO define for DRGs)

**Definition 2.1.3** (Cliques and Cocliques [1, Section 3.3]) Let  $Y \subseteq X$ , and  $C \subseteq D$ , where  $0 \in C$ . Then Y is called a C-CLIQUE if

$$R_i \cap Y^2 = \emptyset \quad \forall i \in D \setminus C .$$

Equivalently, Y is a C-clique if for all  $x, y \in Y$ ,

$$(x,y) \in R_i \implies i \in C$$
.

Let  $C^* := C \setminus \{0\}$ , and  $\overline{C} = D \setminus C^*$ . Then Y is a C-clique if and only if it is a  $\overline{C}$ -coclique. (TODO Move below note to DRG defn.) (Note that such a set Y is called a C-code, or  $\overline{C}$ -anticode in [3].)

### 2.1.1 The Bose-Mesner Algebra

### 2.1.2 Duality

## 2.2 P-Polynomial Schemes

- Definitions
- "Equivalence" to DRGs

If we want to focus on the LP bound and translation schemes, I'm not sure that this section is necessary, but it is very interesting and provides an important class of examples.

### 2.2.1 Q-Polynomial Schemes

### 2.3 Automorphisms and Cayley Graphs

This section may be merged with the following section.

- Action of a regular group of automorphisms
- Cayley graphs
- Eigenspaces from characters

The material in this section follows mostly from [3, Chapter 9].

This section and the next describes a special class of symmetric graphs (respectively, association schemes). As with many other mathematical structures, the *symmetry* of graphs (or schemes) is made precise by examining its automorphisms – those transformations of the object in question which leave its structure unchanged. Graphs (or schemes) with certain automorphisms may be classified in this way.

More importantly for the purposes of this report, the structure revealed by the automorphisms of graphs (or schemes) allows one to compute their eigenvalues significantly more efficiently than otherwise would be the case, as outlined in the previous sections.

### **Definition 2.3.1** (Automorphism)

For graphs  $\Gamma, \Gamma'$ , a map  $\varphi : V(\Gamma) \to V(\Gamma')$  is a HOMOMORPHISM if

$$\forall u, v \in V(\Gamma) \ u \sim_{\Gamma} v \implies \varphi(u) \sim_{\Gamma'} \varphi(v)$$
.

An isomorphism is an invertible homomorphism whose inverse is also a homomorphism; an automorphism is an isomorphism from a graph to itself. Aut  $\Gamma$  denotes the set of all

automorphisms on  $\Gamma$ ; it is the subgroup of Sym  $V(\Gamma)$ , consisting of those permutations which preserve the (edge) structure of  $\Gamma$ .

An automorphism of an association scheme A on vertex set X is a map  $X \to X$  which is simultaneously an automorphism of every class in the scheme. In other words, Aut A is the intersection of the automorphism groups of each class.

### **Definition 2.3.2** (Cayley Graphs)

Given any group G and a subset  $C \subseteq G$ , then C is inverse-closed if for each  $g \in C$ ,  $g^{-1} \in C$  as well.

If  $C \subseteq G$  is an inverse-closed subset of a group G, then the CAYLEY GRAPH of G with respect to C is denoted Cay(G, C), and defined as follows:

- Its vertex set is G
- $g \sim h$  in Cay(G, C) if and only if  $gh^{-1} \in C$

Since C is inverse closed,  $gh^{-1} \in C \iff hg^{-1} \in C$  so that Cay(G, C) is undirected. Furthermore, if  $1_G \notin C$ , then  $g \not\sim g$  so that the graph is loopless. (By definition it already lacks parallel edges.)

Because Cayley graphs are defined from groups using only the group structure, it is intuitive that these graphs should be highly symmetric. For example, every Cayley graph is VERTEX TRANSITIVE: for every pair u, v in the vertex set, there is a automorphism taking  $u \mapsto v$ . To see this, note that G acts (B.1.1) on  $\operatorname{Cay}(G, C)$  through the group operation, since the vertex set is also the group. To verify that this is a homomorphism, if  $u \sim v$  in  $\operatorname{Cay}(G, C)$ , then

$$uv^{-1} \in C \implies (ug)(vg)^{-1} = ugg^{-1}v^{-1} = u1v^{-1} = uv^{-1} \in C$$

so that  $ug \sim vg$ . Then,  $\operatorname{Cay}(G,C)$  is clearly vertex transitive if G acts transitively, and for any vertices u,v, the group element  $u^{-1}v$  maps u to v. Moreover, this action is free, since if ug = u, then the group cancellation law implies that g is trivial. Together, this implies that the action of G is regular, which suggests the following lemma which provides a characterization of Cayley graphs.

(This action is also faithful since for  $g \neq h \in G$ , the vertex 1 gets mapped to g and h respectively, which are unequal. However this observation irrelevant for this lemma, since it restricts to the action of an automorphism group, which is automatically faithful.)

### Lemma 2.3.3

For a graph  $\Gamma$ , there exists a subgroup  $G \leq \operatorname{Aut} \Gamma$  which acts regularly on  $\Gamma$  if and only if  $\Gamma \cong \operatorname{Cay}(G, C)$  for some inverse-closed  $C \subseteq G$ .

Since the above argument demonstrates the reverse implication, only the forward direction will be shown here.

Before beginning the proof, it will be worthwhile to note the neighbours of  $1_G$  in Cay(G, C):  $g \sim 1_G$  precisely when  $g1^{-1} = g \in C$ .

*Proof.* Choose a vertex  $v \in V(\Gamma)$  to identify with  $1_G$ . (This choice will not matter in the end, as Cayley graphs are vertex transitive.) Since the action is regular, for each  $u \in V(\Gamma)$  there exists a unique  $g_u \in G$  such that  $vg_u = u$  (B.1.4).

Then define

$$C := \{ g \in G \mid vg \sim v \}$$

and observe that for  $u, w \in V(\Gamma)$ ,  $ug_u^{-1} = v$ , and  $w = vg_w \implies wg_u^{-1} = vg_wg_u^{-1}$ . So, since  $g_u^{-1}$  is an automorphism of  $\Gamma$ ,

$$u \sim w \iff ug_u^{-1} \sim wg_u^{-1} \iff v \sim vg_wg_u^{-1} \iff g_wg_u^{-1} \in C$$
.

Therefore, the map  $u \mapsto g_u$  is the desired isomorphism  $\Gamma \to \text{Cay}(G, C)$ .

As promised at the beginning of the section, the next lemma demonstrates (for graphs) how automorphisms may be used to derive eigenvalues, and moreover, their eigenvectors. Naively, computing the eigenvalues of a matrix A involves solving its characteristic polynomial, which is generically difficult. Then for an eigenvalue  $\theta$ , finding a  $\theta$ -eigenvector involves computing the kernel of  $A - \theta I$ , which can be computed in polynomial time (though not in linear time), and fast numeric algorithms are typically inexact. (TODO citation)

However, given the right information about a group, the following result finds the eigenvectors and eigenvalues almost instantaneously.

#### Lemma 2.3.4

Let G be a finite abelian group, let  $C \subseteq G \setminus \{1\}$  be inverse-closed, and define  $\Gamma := \operatorname{Cay}(G, C)$ . Then the rows of the character table of G provide a complete set of eigenvectors for the adjacency matrix A of  $\Gamma$ . Specifically, if  $\psi$  is a character of G (equivalently, a row of its character table), then  $\psi(C)$  is the eigenvalue of  $\psi$ .

*Proof.* Note first that the neighbours  $h \sim g$  of a vertex  $g \in G$  consist of precisely the set  $\{cg \mid c \in C\} = Cg$  since  $h \sim g \iff hg^{-1} \in C$ , and multiplication by g is invertible.

As in (B.1), characters are identified with row vectors such that  $\psi(g) \leadsto \psi_g$ . Then

$$(A\psi)_g = \sum_{h \in G} A_{g,h} \psi(h) = \sum_{h \sim g} \psi(h) = \sum_{c \in C} \psi(cg) = \psi(g) \sum_{c \in C} \psi(c) = \psi_g \psi(C) \ .$$

Furthermore, since the rows of the character table are orthogonal, the eigenvectors  $\psi$  are

linearly independent, and since  $G \cong G^*$  (B.2.2) implies that the character table is square, the rows form a basis of eigenvectors for A.

TODO How to get real eigenvectors out of this?

### 2.4 Partitions and Translation Schemes

- Equitable partitions of matrices
- Group partitions yielding association schemes
- Dual schemes? (Interesting, but not particularly necessary for the rest of this report)

In a sense, this section generalizes the characterization of Cayley graphs from the previous section to the setting of association schemes. Throughout this section, a transitive, abelian group of automorphisms will replace the regular automorphism group which corresponds to a Cayley graph. As per (B.1.3) the transitive, abelian group will act regularly, so that (2.3.3) still applies. This motivates the following definition.

### **Definition 2.4.1** (Translation Schemes)

A TRANSLATION SCHEME is an association scheme whose automorphism group contains a transitive, abelian subgroup.

### Lemma 2.4.2

If  $\mathcal{A}$  is a translation scheme, and G is a transitive, abelian automorphism group, then there is a partition into inverse-closed sets  $C_0, C_1, \ldots, C_d$  of G where  $C_0 = \{1\}$ , and each graph  $\Gamma_i$  in  $\mathcal{A}$  is isomorphic to  $Cay(G, C_i)$ .

*Proof.* Since G is abelian and is a transitive subgroup of Aut  $\Gamma_i$  for each i = 0, 1, ..., d, G acts regularly on  $\Gamma_i$ . Therefore, by (2.3.3), there exists an inverse-closed set  $C_i \subseteq G$  such that  $\Gamma_i \cong \text{Cay}(G, C_i)$ .

In particular, since the edges of  $\Gamma_0$  are the diagonal relation,  $C_0 = \{1\}$  generates the graph.

Otherwise, it suffices to show that  $C_0, C_1, \ldots, C_d$  partition G. Recall from the proof of (2.3.3) that any vertex may be chosen to identify with  $1_G$ , so that the same vertex (say, v) may be chosen for each graph  $\Gamma_i$  without loss of generality, in which case  $C_i$  consists of the neighbours of v. By the definition of an association scheme, for each vertex u there is exactly one graph  $\Gamma_i$  in which  $u \sim v$ , so that for each vertex, there is exactly one  $C_i$  containing it.

In order to characterize the translation schemes in a similar manner to the Cayley graphs, an examination of partitions of matrices and groups will be required. This will lead to a simple criterion that distinguishes those partitions which generate a translation scheme translation scheme from those which do not. [3, Section 12.10]

### **Definition 2.4.3** (Partition Matrix)

If  $\sigma$  is a partition of a set X, then the Partition matrix of  $\sigma$  is the 01 matrix whose rows are indexed by the elements of X, and whose columns are indexed by the parts of  $\sigma$ , in which each row – corresponding to  $x \in X$  – has exactly one 1, in the column corresponding to the part that contains x.

Any partition matrix may be obtained from an  $X \times X$  identity matrix by merging the columns which correspond to elements in the same part. Note that this implies that the columns are linearly independent. (The rows will **not** be linearly independent unless the partition is induced by the diagonal relation.)

#### **Definition 2.4.4** (Induced Row Partition)

Given a matrix H, if  $\sigma$  is a partition of the columns with partition matrix  $\chi(\sigma)$  then the INDUCED ROW PARTITION  $\sigma^*$  is the partition of the rows of H such that two rows are in the same part if and only if the corresponding rows in  $H\chi(\sigma)$  are equal.

In other words, if f is the function which maps each row index i of H to the row vector  $(H\chi(\sigma))_i$ , then  $\sigma^*$  is the partition given by the fibres of f. [3, Section 12.7]

### **Theorem 2.4.5** (Bridges and Mena [3, Theorem 12.10.1])

Let G be a finite abelian group, let  $\sigma = \{C_0, C_1, \dots, C_d\}$  be a partition of G into inverseclosed parts where  $C_0 = \{1\}$ , and let  $\sigma^*$  be the induced row partition of the character table H of G.

Then  $|\sigma^*| \geq |\sigma|$ , and the graphs  $\Gamma_i := \operatorname{Cay}(G, C_i)$  form the classes of an association scheme if and only if  $|\sigma^*| = |\sigma|$ .

Proof. Let  $A_i$  be the adjacency matrix of  $\Gamma_i$ , and observe that the set  $\{A_0, A_1, \ldots, A_d\}$  is linearly independent. This is because the sets  $C_i$  partition G, and in each  $\Gamma_i$  the set  $C_i$  consists of the neighbours of 1. The fact that the  $C_i$  partition G also implies that  $\sum_i A_i = J$ , and since  $C_0 = \{1\}$ ,  $A_0 = I$ .

By (2.3.4), each character  $\psi$  of G (i.e. row of H) is a common eigenvector of  $A_0, A_1, \ldots, A_d$ , with eigenvalue  $\psi(C_i)$  at  $A_i$ . Define  $\mathbb{A} := \operatorname{span} \{A_0, A_1, \ldots, A_d\}$ . Let  $\chi_{C_i}$  be the characteristic vector of  $C_i$  in G, and let

$$\chi(\sigma) = \begin{bmatrix} | & | & | \\ \chi_{C_0} & \chi_{C_1} & \cdots & \chi_{C_d} \\ | & | & | \end{bmatrix}$$
 (2.3)

be the partition matrix of  $\sigma$ .

Let  $D_0, D_1, \ldots, D_e$  be the parts of  $\sigma^*$ ; then i, k (or, their characters  $\psi^i, \psi^k$ ) belong to the same part  $D_i$  precisely when the rows  $\psi^i \chi(\sigma), \psi^k \chi(\sigma)$  in

$$H\chi(\sigma) = \begin{bmatrix} - & \psi^1 & - \\ & \vdots & \\ - & \psi^n & - \end{bmatrix} \begin{bmatrix} | & | & | \\ \chi_{C_0} & \chi_{C_1} & \cdots & \chi_{C_d} \\ | & | & | \end{bmatrix}$$
 (2.4)

are equal. Together, the characters of each  $D_j$  span a common eigenspace of the  $A_i$ : let  $F_j$  be the orthogonal projection matrix onto this subspace.

Define  $\mathbb{F} := \operatorname{span} \{F_0, F_1, \dots, F_e\}$ . Since the  $\operatorname{col} F_j$  are spanned by disjoint sets of characters, the subspaces are orthogonal and the  $F_j$  are linearly independent; since together the characters span  $\mathbb{C}^n$  (where n is the order of G), the (direct) sum of the subspaces is  $\mathbb{C}^n$  as well. Therefore,

$$I = F_0 + F_1 + \dots + F_e .$$

Furthermore, since  $\operatorname{col} F_j$  is a common eigenspace for the  $A_i$ , for  $i = 0, 1, \ldots, d$  and  $j = 0, 1, \ldots, e$  there exist constants  $P_i(j)$  such that

$$A_i F_j = P_i(j) F_j \implies A_i = \sum_{j=0}^e P_i(j) F_j \implies \mathbb{A} \le \mathbb{F}$$
 (2.5)

This implies that

$$|\sigma| = d = \dim \mathbb{A} \le \dim \mathbb{F} = e = |\sigma^*|$$
.

Note that the  $F_0, F_1, \ldots, F_e$  are orthogonal idempotents, so they are closed under the regular matrix product. This implies that the algebra they generate is simply  $\mathbb{F}$ . On the other hand, while  $A_0, A_1, \ldots, A_d$  are orthogonal idempotents with respect to the *Schur product*, they may generate an algebra with the usual product that is strictly larger than  $\mathbb{A}$  – it must, however, be contained in  $\mathbb{F}$ . We will show that these two algebras are actually equal. In this case, e = d if and only if  $\mathbb{A}$  is closed under regular matrix multiplication; given the results above, this will then be true if and only if  $A_0, A_1, \ldots, A_d$  forms an association scheme.

From (TODO reference) and (2.5), if g(x) is any polynomial, then  $g(A_i) = \sum_j g(P_i(j)) F_j$ . In particular, if  $x_0^{s_0} x_1^{s_1} \cdots x_d^{s_d}$  is any monomial,  $A_i^{s_i} = \sum_j P_i(j)^{s_i} F_j$  as above, so that evaluating the monomial at  $(A_0, A_1, \ldots, A_d)$  yields

$$A_0^{s_0} A_1^{s_1} \cdots A_d^{s_d} = \prod_i \sum_j P_i(j)^{s_i} F_j = \sum_j \left( \prod_i P_i(j)^{s_i} \right) F_j$$

since the  $F_j$  are orthogonal idempotents. Since any polynomial g in d+1 variables is a

linear combination of such monomials, it follows that

$$g(A_0, A_1, \dots, A_d) = \sum_{j} g(P_0(j), P_1(j), \dots, P_d(j)) F_j$$

$$\implies g(A_0, A_1, \dots, A_d) F_j = g(P_0(j), P_1(j), \dots, P_d(j)) F_j$$

for all j = 0, 1, ..., e.

Now let P be the  $(e+1) \times (d+1)$  matrix such that  $P_{ji} = P_i(j)$ . Note that the rows of P are precisely the distinct rows of  $H_{\chi}(\sigma)$ , so that for any two rows  $j \neq j'$  of P, there exists a column i(j,j') such that  $P_{i(j,j')}(j) \neq P_{i(j,j')}(j')$ . This allows for the definition of the polynomials

$$g_j(x_0, x_1, \dots, x_d) := \prod_{j' \neq j} (x_{i(j,j')} - P_{i(j,j')}(j'))$$

so that when applied at  $A_0, A_1, \ldots, A_d$ ,

$$g_j(A_0, A_1, \dots, A_d) F_{j''} := \prod_{j' \neq j} \left( P_{i(j,j')}(j'') - P_{i(j,j')}(j') \right) F_{j''}$$
$$= g_j \left( P_0(j''), P_1(j''), \dots, P_d(j'') \right) F_{j''}.$$

By construction, if j''=j, then  $f_j:=g_j\left(P_0(j''),P_1(j''),\ldots,P_d(j'')\right)$  will be non-zero, but if  $j''\neq j$ , then there will be some j'=j'' at which  $P_{i(j,j')}(j'')-P_{i(j,j')}(j')=0$  so that  $g_j\left(P_0(j''),P_1(j''),\ldots,P_d(j'')\right)=0$ .

This construction demonstrates that for each j, there exists a polynomial  $g_j$  such that

$$g_j(A_0, A_1, \dots, A_d) = \sum_{j'} f_j \delta_{jj'} F_{j'} = f_j F_j$$

where  $f_j \neq 0$ , so that  $F_j$  can be written as a polynomial in  $A_0, A_1, \ldots, A_d$ . This proves that each  $F_j$  is contained in the algebra generated by  $\mathbb{A}$ , and so all of  $\mathbb{F}$  is contained in this algebra. Since the reverse inclusion was already shown, this proves that the two algebras are equal, as desired.

It is interesting to note that, while the character table H may in general be complex, each of the orthogonal projection matrices  $F_j$  is real. To see this, note that the  $A_i$  are real, symmetric matrices, so that their eigenvalues  $P_i(j)$  are real as well. Then, each of the polynomials  $g_j$  (used to express  $F_j$  in the algebra generated by the  $A_i$ ) must also be real, since they were defined in terms of the  $P_i(j)$ . Therefore, not only are the  $\mathbb{C}$ -algebras of  $\mathbb{A}$  and  $\mathbb{F}$  equal, but so are their  $\mathbb{R}$ -algebras.

With this result, translation schemes are characterized by a finite abelian group G and partition  $\sigma$  satisfying the condition given. Moreover, a finite abelian group G is isomorphic to its groups of characters,  $G^*$  (B.2.2), and the group of characters is completely described by the character table H. Likewise, the group partition  $\sigma$  is completely described by its partition matrix  $\chi(\sigma)$ . Since the condition in (2.4.5) depends only on these two matrices (both with respect to the same ordering on G), if it is satisfied, then the matrices completely describe the translation scheme they generate.

# 2.5 The Eigenvalues of the Hamming Scheme

The theory just developed in the previous section can be applied immediately to the Hamming scheme. This scheme is of great utility in the setting of coding theory, in part because it is a *P*-polynomial scheme, generated by the distance-regular Hamming graph. Moreover, it is also a translation scheme, with respect to a particularly nice group, and a simple partition, which will allow us to deduce a formula for the eigenvalues of the scheme. In the particular case of the Hamming graph, an explicit expression for its eigenvalues can be given.

To this end, let  $\mathbb{Z}_q$  denote the cyclic group of order q (written additively), and consider the direct product  $\mathbb{Z}_q^d$  with subsets

$$C_i := \left\{ x \in \mathbb{Z}_q^d \mid \text{there are exactly } i \text{ 0's in } x \right\}$$
.

In particular,  $C_0 = \{0\}$  and

$$C_1 = \mathbb{Z}_q \setminus \{0\} \times \{0\}^{d-1} \mid \{0\} \times \mathbb{Z}_q \setminus \{0\} \times \{0\}^{d-2} \mid \cdots \mid \{0\}^{d-1} \times \mathbb{Z}_q \setminus \{0\}$$
.

Then x, y are  $i^{\text{th}}$  associates if and only if  $x - y \in C_i$ ; that is, the Hamming distance between x and y is i, so that this partition yields the Hamming scheme. In particular,  $H(d,q) \cong \text{Cay}(\mathbb{Z}_q^d, C_1)$ .

For a character of the group  $\psi \in (\mathbb{Z}_q^d)^*$  (B.2.1) and  $x = (x_1, \dots, x_d) = \sum_{i=1}^d x_i e_i$  in the group (here  $e_i$  is the tuple of all zeroes, and a 1 in the  $i^{\text{th}}$  spot), then since  $\psi$  is a homomorphism,

$$\psi(x) = \prod_{i=1}^{d} \psi(x_i e_i) = \prod_{i=1}^{d} \psi\left(\sum_{j=1}^{x_i} e_i\right) = \prod_{i=1}^{d} \prod_{j=1}^{x_i} \psi(e_i) = \prod_{i=1}^{d} \psi(e_i)^{x_i}$$

so that  $\psi$  is completely determined by its values on  $e_1, \ldots, e_d$ .

Let  $\omega$  be a primitive  $q^{\text{th}}$  root of unity, so that  $1 = \omega^0, \omega^1, \dots, \omega^{q-1}$  are distinct. Then every choice in  $\{\omega^0, \omega^1, \dots, \omega^{q-1}\}^d$  will yield a distinct character by assigning the  $i^{\text{th}}$  entry

to  $\psi(e_i)$  (identifying  $\psi$  with the tuple as in (B.1)), and defining the value of  $\psi$  at all other  $x = (x_1, \ldots, x_d)$  by

$$\psi(x) := \prod_{i=1}^d \psi(e_i)^{x_i} .$$

Note that this is a homomorphism since

$$\psi(x+y) = \prod_{i=1}^d \psi(e_i)^{x_i+y_i} = \prod_{i=1}^d \psi(e_i)^{x_i} \psi(e_i)^{y_i} = \prod_{i=1}^d \psi(e_i)^{x_i} \prod_{i=1}^d \psi(e_i)^{y_i} = \psi(x)\psi(y) .$$

Therefore,  $(\mathbb{Z}_q^d)^* \cong \{\omega^0, \dots, \omega^{q-1}\}^n$  (taking entrywise multiplication as the group product on the right), so that the characters  $\psi$  will be identified with row vectors

$$\psi \leadsto \begin{bmatrix} \omega^{\psi_1} & \cdots & \omega^{\psi_d} \end{bmatrix}$$

where  $\psi_i \in \{0, 1, \dots, q-1\}.$ 

(Note that this notation deviates from (B.1), but will be more convenient for this purpose. In fact, this shows that  $(\mathbb{Z}_q^d)^* \cong \mathbb{Z}_q^d$  directly, confirming (B.2.2).)

Then, in the  $i^{\text{th}}$ -distance Hamming graph,  $\psi$  is a  $\psi(C_i)$ -eigenvalue (2.3.4), and for the Hamming graph, it can be computed directly.

$$\psi(C_1) = \sum_{c \in C_1} \psi(c)$$

$$= \sum_{i=1}^d \sum_{j=1}^{q-1} \psi(je_i) \quad \text{here, the outer sum picks which entry of } c \text{ will be non-zero}$$

and the inner sum picks the value

$$= \sum_{i=1}^{d} \sum_{j=1}^{q-1} \psi(e_i)^j = \sum_{i=1}^{d} \sum_{j=1}^{q-1} \left(\omega^{\psi_i}\right)^j$$

$$= \sum_{i=1}^{d} \left(\frac{1 - \left(\omega^{\psi_i}\right)^q}{1 - \omega^{\psi_i}} - 1\right) \quad \text{using the usual formula for geometric sums}$$

$$= \sum_{i=1}^{d} (q-1)\delta_{0,\psi_i} - d$$

$$= (q-1) \text{ (the number of } i \text{ with } \psi_i = 0) - d$$

Since the number k of indices i for which  $\psi_i = 0$  can vary from  $0, 1, \ldots, d$ , and there are  $\binom{d}{k}$  places i at which  $\psi_i = 0$  and  $(q-1)^{d-k}$  choices for the other  $\psi_j$ , the eigenvalues of the

Hamming graph are given

$$\begin{cases} qk - d & k = 0, 1, \dots, d \\ {d \choose k} (q - 1)^{d - k} & \text{(multiplicy)} \end{cases}$$
 (2.6)

For the other graphs in the Hamming scheme, their eigenvalues can be computed as  $\psi(C_i)$ , although there may not be a particularly nice expression for this sum.

# 3. Delsarte's Linear Programming Bound

### 3.1 Linear Programming

- Basics of Linear Programming done
- Duality done
- Algorithms?

The terminology and results from this section, except for the adjective *principal* for constraints, follows from [4].

A LINEAR PROGRAMMING PROBLEM (or LINEAR PROGRAM) is an optimization problem in which one seeks to maximise or minimize a linear function of one or more variables, subject to linear constraints. That is, fixing a vector c, one tries to maximize or minimize the linear combinations of the components of c:

$$c_1 x_1 + \dots + c_n x_n = c^T x$$

for some x. Note that maximizing  $c^T x$  is equivalent to minimizing  $(-c)^T x$ , so that for the theory of linear programming, it suffices to consider maximization problems without loss of generality. As in other optimization problems, the function to be maximized  $(c^T x)$  in this case) is called the OBJECTIVE (FUNCTION).

In most cases, there will be contraints on the inputs to the objective function, and for the purposes of linear programming these will also have to be linear. That is, there will be a matrix A and vector b such that only inputs x satisfying  $Ax \leq b$  will be allowed. (Note that for vectors a and b,  $a \leq b$  will mean that each component  $a_i$  is less than or equal to the corresponding component  $b_i$ .) These are called the (PRINCIPAL) CONSTRAINTS, and vectors x which satisfy the constraints will be called FEASIBLE (SOLUTIONS). (Note that in [1], the term program is used to refer to a feasible solution.)

If there are no constraints on the problem (and even in some cases where there are) through appropriate choices of feasible solution x, the objective  $c^T x$  may be made arbitrarily

large, and such problems are called UNBOUNDED. Conversely, if no feasible solutions exist, then the problem is called INFEASIBLE.

Finally, in most applications of linear programming – in particular to the cliques of association schemes – the feasible solutions will be further constrained to those with all nonnegative components (i.e.  $x \geq 0$ ). These are called the NON-NEGATIVITY CONSTRAINTS, in constrast with the *principal constraints*. The non-negativity constraints will be required throughout the remainder of this report.

Therefore, for an objective  $c^T x$  and constraints  $Ax \leq b$ , the associated linear program will be written in STANDARD FORM:

$$\max\left\{c^T x \mid Ax \le b, \ x \ge 0\right\} .$$

### 3.1.1 Duality

The most important observation about linear programs (for the purposes of this report, at least) is that they come in dual pairs.

Given a linear program  $\mathcal{P}$  written in standard form

$$\max\left\{c^T x \mid Ax \le b, \ x \ge 0\right\}$$

its DUAL program is  $\mathcal{P}^*$ :

$$\min\left\{b^Ty\ \big|\ A^Ty\geq c,\ y\geq 0\right\}\ .$$

Re-writing it in standard form,

$$\max\left\{(-b)^Ty\ \big|\ -A^Ty \le -c,\ y \ge 0\right\}$$

taking the dual

$$\min\left\{-c^T x \mid -Ax \ge -b, \ x \ge 0\right\}$$

and re-writing in standard form

$$\max \left\{ c^T x \mid Ax \le b, \ x \ge 0 \right\}$$

the original (called PRIMAL) linear program is recovered.

This demonstrates that  $(\mathcal{P}^*)^* = \mathcal{P}$ , so that linear programs come in dual pairs.

### **Theorem 3.1.1** (Weak Duality)

If x is a feasible solution to a linear program

$$\max\left\{c^T x \mid Ax \le b, \ x \ge 0\right\},\$$

and y is a feasible solution to its dual program,

$$\min \left\{ b^T y \mid A^T y \ge c, \ y \ge 0 \right\},\,$$

then  $c^T x \leq b^T y$ .

*Proof.* Let u, v, w be vectors with  $u \ge 0$ , and  $v \le w$ . Then for all components  $i, u_i \ge 0$  and  $v_i \le w_i$  implies that  $u_i v_i \le u_i w_i$  so that

$$u^T v = \sum_i u_i v_i \le \sum_i u_i w_i = u^T w .$$

In particular, since y is a feasible solution to the dual program, and  $x \ge 0$ ,

$$c < A^T y \implies x^T c < x^T A^T y = y^T A x$$
.

(Here one may take the transpose of the whole expression, since the result is a scalar.) Similarly, since x is a feasible solution to the primal program, and  $y \ge 0$ ,

$$b \ge Ax \implies y^T b \ge y^T Ax$$
.

By combining the two inequalities,

$$b^T y = y^T b \ge y^T A x \ge x^T c = c^T x$$

which is the desired result.

As a result of the weak duality of linear programs, every feasible solution to the dual program provides an upper bound on the maximum of the primal, and every feasible solution to the primal program provides a lower bound on the minimum of the dual.

In fact the extremal values of dual programs (the maximum of the primal, and the minimum of the dual) coincide, although this will not be needed for the purposes of this report. This is referred to as *Strong Duality* of linear programs.

### 3.2 The LP Bound

### Definition 3.2.1

The INNER DISTRIBUTION is TODO. I might also put this in the section on association schemes; I'm not sure if it belongs better there or here.

**Theorem 3.2.2** (Delsarte Thm 3.3 [1])

For any inner distribution y,

$$Q^T y \ge 0$$

where Q is the matrix of dual eigenvalues. (Here,  $x \ge 0$  means that each component of the vector x is not less than 0.)

*Proof.* TODO. Note that this will require a number of lemmas which I've omitted here for brevity, but will include in the final product.  $\Box$ 

This theorem provides the key inequality that will allow the application of linear programming to cliques in association schemes. However, because the constraint vector in a primal linear program becomes the objective in the dual program, this inequality will require some transformation to make it suitable for use in linear programming.

Let Y be an M-clique with inner distribution y. Then  $y_i = 0$  for all  $i \notin M$ , so  $Q^T y \ge 0 \iff Q^T \operatorname{diag}(\chi_M) y \ge 0$  since the action of  $\operatorname{diag}(\chi_M)$  acting on the left is to zero out the rows of y with index not in M. Similarly,

$$Q^{T} \operatorname{diag}(\chi_{M}) y = Q(0)^{T} y_{0} + Q^{T} \operatorname{diag}(\chi_{M^{*}}) y = \mu + Q^{T} \operatorname{diag}(\chi_{M^{*}}) y$$

since the action of diag  $(\chi_{M^*})$  on the right is to zero out the *columns* of  $Q^T$  with index not in  $M^*$ ,  $y_0 = 1$ , and

$$Q^T = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ \mu_1 & & & \\ \vdots & & * & \\ \mu_d & & & \end{bmatrix} .$$

Finally, since  $y \ge 0$ ,  $Q_0^T y \ge 0$  adds no new constraint, so that under the non-negativity constraint  $Q^T y \ge 0 \iff \operatorname{diag}(\chi_{N^*}) Q^T y \ge 0$ .

Putting all this together,  $Q^T y \geq 0 \iff \operatorname{diag}(\chi_{N^*}) Q^T \operatorname{diag}(\chi_{M^*}) y \geq -\mu$  so that Delsarte's LP can be written in standard form:

$$\max \{ \mathbf{1}^T \operatorname{diag}(\chi_M) y \mid Q^T \operatorname{diag}(\chi_M) y \ge 0, \ y \ge 0, \ y_0 = 1 \}$$
(3.1)

$$= \max \left\{ \chi_{M^*}^T y \mid -\operatorname{diag}(\chi_{N^*}) Q^T \operatorname{diag}(\chi_{M^*}) y \le \operatorname{diag}(\chi_{N^*}) \mu, \ y \ge 0 \right\} + 1 \ . \tag{3.2}$$

Taking the dual yields

$$\min \left\{ \mu^T \operatorname{diag}(\chi_{N^*}) z \mid -\operatorname{diag}(\chi_{M^*}) Q \operatorname{diag}(\chi_{N^*}) z \ge \chi_{M^*}, \ z \ge 0 \right\} + 1$$
 (3.3)

$$= \min \left\{ \mu^T z \mid -\operatorname{diag}(\chi_{M^*}) Q \operatorname{diag}(\chi_{N^*}) z \ge \chi_{M^*}, \ z \ge 0, \ z_0 = 1 \right\} . \tag{3.4}$$

Therefore, if  $z_0 = 1$  is required, recalling that  $Q_0 = \mathbf{1}$  and diag  $(\chi_{M^*}) \mathbf{1} = \chi_{M^*}$ , then

$$\begin{aligned} &\operatorname{diag}\left(\chi_{M^*}\right) Q \operatorname{diag}\left(\chi_{N}\right) z \\ &= \operatorname{diag}\left(\chi_{M^*}\right) \left(Q_0 z_0 + Q \operatorname{diag}\left(\chi_{N^*}\right) z\right) \\ &= &\chi_{M^*} + \operatorname{diag}\left(\chi_{M^*}\right) Q \operatorname{diag}\left(\chi_{N^*}\right) z \\ &\leq &0 \ . \end{aligned}$$

This equivalence recover's Delsarte's formulation of the dual linear program:

$$\min \left\{ \mu^T z \mid \operatorname{diag}(\chi_{M^*}) \, Qz \le 0, \ z \ge 0, \ z_0 = 1 \right\} . \tag{3.5}$$

## 3.3 The Ratio Bound

We will use (3.5) frequently.

# 3.4 The Clique-Coclique Bound

# 4. Schrijver's SDP Bound

- 4.1 The Terwilliger Algebra of the Hamming Scheme
- 4.2 Semi-Definite Programming

# 5. Computation

I wasn't sure if I ought to mention anything about the code I've written for this project (or even if there's anything worth saying that won't be covered elsewhere in the report).

Also, if there are some specific results that would be interesting to show, but do not fit naturally into other sections of the report, then perhaps they could go here as well.

• Computing the character table of an abelian group

# 6. Discussion

I'm not sure if "Discussion" is the right name for a chapter of this sort, if it is even worth including. If it is, I would try and keep this part brief.

### 6.1 Conclusion

I know papers typically have some sort of conclusion or summary towards the end, but I wasn't sure if it would be valuable to include something like that in a report of this kind.

# 6.2 Other Applications

Perhaps it might be worth mentioning number of other applications of association schemes, for example to design theory, or statistics?

# A. Linear Algebra

# A.1 The Spectral Theorem

# A.2 Adjacency Matrices

Basic results about the spectra of adjacency matrices, which may be used elsewhere in the report. E.g. the sum of eigenvalues with multiplicity, and consequences.

# A.3 Positive Semi-Definite Matrices

Depending on which proof of the clique-coclique bound I use, and how much detail I go into Schrijver's SDP bound, I could make some comments about PSD matrices.

# B. Group Theory

# **B.1** Group Actions

The material of this section comes primarily from [2, Section 1.7, Chapter 4].

### **Definition B.1.1** (Group Action)

Given a group G and a set X, GROUP ACTION is a homomorphism  $G \to \operatorname{Sym} X$ , where  $\operatorname{Sym} X$  is the symmetric group on X.

A group action  $\varphi: G \to \operatorname{Sym} X$  induces a product  $X \times G \to X$  by mapping  $(x,g) \mapsto \varphi(g)(x)$ . When the action is clear from context, this will be denoted  $x \cdot g$ , or simply xg. This is called a RIGHT ACTION, as g acts on the right of x (the corresponding notion of a LEFT ACTION can also be defined.)

Conversely, given a product  $X \times G \to X$ , the same expression defines a map  $G \to \operatorname{Sym} X$ . If such a product satisfies

$$\forall x \in X \ x1_G = x$$
 and 
$$\forall x \in X \ \forall g, h \in G \ (xg)h = x(gh)$$

then the induced map  $G \to \operatorname{Sym} X$  will be a homomorphism, so that these definitions are equivalent.

(In [2] this is taken as the definition of a group action, and the homomorphism  $G \to \operatorname{Sym} X$  is called its PERMUTATION REPRESENTATION. It will be occasionally convenient to adopt each perspective.)

### **Definition B.1.2** (Types of Group Actions)

If if a homomorphism  $G \to \operatorname{Sym} X$  is injective, then the action is called faithful. Note that a group homomorphism is injective if and only if it has a trivial kernel.

Given a group action  $G \to \operatorname{Sym} X$ ,  $g \in G$  is called fixed point-free if  $\forall x \in X$   $xg \neq x$ . The group action itself is called fixed point-free (or just free) if all its nontrivial elements are fixed point-free. A group action  $G \to \operatorname{Sym} X$  is called Transitive if  $\forall x, y \in X$  there exists some  $g \in G$  such that xg = y.

A group action is called REGULAR if it is simultaneously transitive and free. (This terminology follows [3].)

Note that if X is a structure with automorphisms (such as a graph or group), G is a subgroup of Aut X, and G acts in the natural way on X (i.e. xg = g(x)), then this action is faithful. That is, Aut  $X \leq \operatorname{Sym} X$ , so that this action is induced by the identity  $G \hookrightarrow \operatorname{Sym} X$ , which is clearly injective.

### Lemma B.1.3

If an abelian group G acts faithfully and transitively on a set X, then the action is free, and thus also regular. [2, Section 4.1, Exercise 3]

*Proof.* Let  $g \in G$  be nontrivial, and  $x \in X$ . The goal is to prove that  $xg \neq x$ .

Since g is not the identity, there exists some  $y \in X$  such that  $z := yg \neq y$ . Furthermore, since G acts transitively on X, there exists some  $h \in G$  such that  $yh = x \iff y = xh^{-1}$ . Then,

$$xg = (yh)g$$
  
 $= y(hg)$   
 $= y(gh)$  since  $G$  is abelian  
 $= (yg)h$   
 $= zh$ .

If zh = x then,  $z = xh^{-1} = y$ , but by definition,  $z = yg \neq y$ , so  $xg = zh \neq x$ .

An alternate characterization of regular actions will be useful in this report. To see this, note that for a pair  $x, y \in X$ , there exists a  $g \in G$  such that xg = y by transitivity; for any  $g' \in G$  satisfying xg' = y,

$$xg = xg' \implies x = xg'g^{-1}$$

so that  $g'g^{-1} = 1$  since the action is free, and so g' = g. Conversely, if for each  $x, y \in X$  there existed a unique  $g \in G$  satisfying xg = y, then the action would clearly be transitive; since x1 = x, 1 is the unique group element fixing any point, so the action must be free.

### Lemma B.1.4

A group action G on X is regular if and only if for all  $x, y \in X$ , there exists a unique  $g \in G$  such that xg = y.

### B.2 Character Theory

### **Definition B.2.1** (Characters)

Given a group G, a CHARACTER of the group G is a homomorphism  $G \to \mathbb{C}^{\times}$ , the group of non-zero complex numbers under multiplication. Then  $G^*$  will denote the set of characters of G. [3, Chapter 8]

(For abelian groups, this corresponds to irreducible degree 1 characters over  $\mathbb{C}$  in [2, Section 18.3].)

TODO Maybe use  $\circ$  for this product to mimic the usage for the Schur product? On  $G^*$  a product of characters can be defined by setting

$$\varphi \psi : g \mapsto \varphi(g)\psi(g)$$

under which the character taking each  $g \in G$  identically to 1 acts as identity.

Furthermore, for any character  $\psi \in G^*$ , the map  $g \mapsto \psi(g^{-1})$  is a homomorphism since

$$gh \longmapsto \psi((gh)^{-1}) = \psi(h^{-1}g^{-1}) = \psi(h^{-1})\psi(g^{-1}) = \psi(g^{-1})\psi(h^{-1})$$

and for any  $g \in G$ ,

$$\psi(g)\psi(g^{-1}) = \psi(g^{-1})\psi(g) = \psi(1) = 1$$

so that this homomorphism is an inverse for  $\psi$ .

Therefore,  $G^*$  forms a group under the above product of characters.

For the remainder of this section (and the rest of this report), discussion of characters will be restricted to the case of finite abelian groups. Throughout this section, G will denote a finite abelian group of order n.

In this case, by Lagrange's theorem,  $g^n=1_G$  for every  $g\in G$ , and so for any character  $\psi\in G^*$ 

$$1 = \psi(1) = \psi(g^n) = \psi(g)^n$$

– that is, the image of each character is contained in the set of  $n^{\text{th}}$  roots of unity.

Since the inverse of a complex number with modulus 1 is also its complex conjugate, looking at the inversion in  $G^*$ ,

$$\psi\left(g^{-1}\right) = \psi(g)^{-1} = \overline{\psi(g)}$$

so that the inverse of  $\psi \in G^*$  is  $\overline{\psi} : g \mapsto \overline{\psi(g)}$ .

### Theorem B.2.2

For all finite abelian groups G,

$$G \cong G^*$$
.

While by transitivity this shows that  $G^{**} \cong G$ , this can be seen more directly via the isomorphism

$$g \mapsto (\psi \mapsto \psi(g))$$
.

Given an ordering  $G = \{g_1, \ldots, g_n\}$ , the character  $\psi \in G^*$  can be identified with the row vector

$$\psi \leadsto \left[ \psi(g_1) \quad \cdots \quad \psi(g_n) \right] .$$
 (B.1)

With this identification, the product of characters becomes the entrywise product of vectors (the *Schur product* of  $n \times 1$  matrices), and the inverse of a character in  $G^*$  is the entrywise inversion of the vector.

Furthermore, given an ordering  $G^* = \{\psi^1, \dots, \psi^n\}$ , the matrix whose rows consist of the characters of  $G^*$  is called the CHARACTER TABLE of G:

$$H = \begin{bmatrix} - & \psi^1 & - \\ & \vdots & \\ - & \psi^n & - \end{bmatrix} . \tag{B.2}$$

Remarkably, this matrix turns out to be (almost) unitary.

For any subset  $C \subseteq G$ , define

$$\psi(C) := \sum_{g \in C} \psi(g) = \psi \chi_C$$

where  $\chi_C$  is the characteristic vector of C in G (with the same ordering). So, given characters  $\psi, \varphi$ , their inner product can be written

$$\psi \varphi^* = \sum_{g \in G} \psi(g) \overline{\varphi}(g) = \sum_{g \in G} (\psi \overline{\varphi})(g) = (\psi \overline{\varphi}) \chi_G = (\psi \overline{\varphi})(G) .$$

(Note here that  $\varphi^*$  denotes the conjugate transpose of  $\varphi$ , and  $\chi_G$  is also the all-ones column vector  $\mathbf{1}$ .)

### Lemma B.2.3

For any  $\psi \in G^*$ 

$$\psi(G) = \begin{cases} |G| & \text{if } \psi \text{ is the identity of } G^* \\ 0 & \text{else.} \end{cases}$$

*Proof.* For any  $h \in G$ , since  $g \mapsto hg$  is an automorphism of G, hG = G, so that

$$\psi(G) = \sum_{g \in G} \psi(g) = \sum_{g \in G} \psi(hg) = \sum_{g \in G} \psi(h) \psi(g) = \psi(h) \sum_{g \in G} \psi(g) = \psi(h) \psi(G)$$

which implies that either  $\psi(h) = 1$  or  $\psi(G) = 0$ .

But this holds for arbitrary  $h \in G$ , so that either

$$\forall h \in G \ \psi(h) = 1 \implies \psi = 1_{G^*} \quad \text{and} \quad \psi(G) = \sum_{g \in G} 1 = |G|$$

or else

$$\exists h \in G \ \psi(h) \neq 1 \implies \psi(G) = 0$$
.

### Corollary B.2.4

If H is the character table of a finite abelian group G of order n, then  $HH^* = nI$ , where  $H^*$  is the conjugate transpose of H.

*Proof.* If  $\psi, \varphi$  are characters of G, and  $\psi \neq \varphi$ , then letting  $\theta = \psi \overline{\varphi}$ , the  $(\psi, \varphi)$ -entry of  $HH^*$  is given by  $\psi \varphi^* = \theta(G) = 0$ , since  $\theta$  is not the identity of  $G^*$ .

However, for the diagonal,  $(\psi, \psi)$ -entries of  $HH^*$ ,  $\psi\psi^* = 1_{G^*}(G) = n$ , which proves the claim.

# B.3 The Structure of Finite Abelian Groups

TODO I need a citation for this. I'm also not sure if this section shouldn't go before the section on Character Theory, as it will be used in (B.2.2).

### **Theorem B.3.1** (Structure Theorem)

If G is a finitely generated abelian group, then G is isomorphic to a direct product of cyclic groups. Specifically,

$$G \cong \mathbb{Z}_{q_1}^{d_1} \times \dots \times \mathbb{Z}_{q_t}^{d_t} \times \mathbb{Z}^f$$

where the  $q_i$  are distinct prime powers. Moreover, this decomposition is unique up to the ordering of its factors.

This is a well-known and well-used result. It comes as a direct result of the Struc-

ture Theorem for Finitely Generated Modules over PIDs, using the language of rings and modules, but this is out of the scope of this report.

# C. Notation

I've included these sections mostly as an excuse to add the citations to the bibliography, though it may be somewhat useful to have a sort of guide to the similarities and differences in notation in this report, as well as in the references. (At least, I would find such a thing useful.)

# C.1 Godsil

See [3] Ch. 10.

### C.2 Delsarte

See [1] Ch. 3.

# C.3 Schrijver

See [5].

# Bibliography

- [1] P. Delsarte. "An algebraic approach to the association schemes of coding theory". PhD thesis. 1973.
- [2] David Steven. Dummit. *Abstract algebra*. eng. 3rd ed. New York: Wiley, 2004. ISBN: 0471433349.
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- [4] Jiří Matoušek. *Understanding and using linear programming*. eng. Universitext. Berlin; Springer, 2007.
- [5] Alexander Schrijver. "New Code Upper Bounds From the Terwilliger Algebra and Semidefinite Programming". In: *Information Theory, IEEE Transactions on* 51 (Sept. 2005), pp. 2859–2866. DOI: 10.1109/TIT.2005.851748.