

Permutation Invariant 2-Matrix Model

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Tensor Journal Club

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Experimental motivation

- Distributional semantics [Harris 1954]
 - Meaning represented by vectors of co-occurrence

Frequency of box co-
 $(\text{box})_i =$ occurring with the i th context
word

- Successful for words, compositional model needed for more complex semantical structures
- Compositional distributional semantics [Coecke, Sadrzadeh, Clark '10]
 - Tensors act on vectors to form sentences
 - Adjectives act on nouns to give *noun phrases*

$$(\text{red})_{ij}(\text{box})_j = (\text{red box})_i$$

- Two matrices give *adjective verb combinations*

$$(\text{eat})_{ij}(\text{quickly})_{jk}(\text{cats})_k = (\text{cats eat quickly})_i$$

Experimental motivation

- Statistics of word matrices should be analyzed using permutation invariant matrix theories [Kartsaklis, Ramgoolam, Sadrzadeh '17]
- 5-parameter permutation invariant Gaussian model showed promising results
- General 13-parameter permutation invariant Gaussian 1-matrix model was solved [Ramgoolam '18]
- Expectation values in good agreement with experiment [Ramgoolam, Sadrzadeh, Sword '19]

1-Matrix Model

[Ramgoolam '18]

- Matrix model with D^2 matrix M_{ij}

$$Z = \int dM e^{-S(M)}$$

- Free (Gaussian)

$$S(M) \sim M + MM$$

- Permutation invariant

$$S(M_{\sigma(i)\sigma(j)}) = S(M_{ij}), \quad \sigma \in S_D$$

- Observables are permutation invariant polynomials

$$f(M_{\sigma(i)\sigma(j)}) = f(M_{ij})$$

1-Matrix Model

[Ramgoolam '18]

- Permutations act on $V_D = \text{Span}(|i\rangle \mid i = 1, \dots, D)$

$$\sigma |i\rangle = |\sigma^{-1}(i)\rangle, \quad \sigma \in S_D$$

- V_D decomposes into irreducible representations

$$V_D \rightarrow V_0 \oplus V_H$$

- There is an orthonormal basis for $V_0 \oplus V_H$

$$|E_0\rangle = \sum_{i=1}^D C_{0i} |i\rangle = \frac{1}{\sqrt{D}} \sum_{i=1}^D |i\rangle$$

$$|E_a\rangle = \sum_{i=1}^D C_{ai} |i\rangle, \quad a = 1, \dots, D-1$$

1-Matrix Model

[Ramgoolam '18]

- Linear combinations of M_{ij} form a vector space $V_D \otimes V_D$

$$V_D \otimes V_D = \text{Span}(|i\rangle \otimes |j\rangle \mid i, j = 1, \dots, D)$$

- S_D acts diagonally on $V_D \otimes V_D$

$$\sigma(|i\rangle \otimes |j\rangle) = \sigma |i\rangle \otimes \sigma |j\rangle$$

1-Matrix Model

[Ramgoolam '18]

- Linear combinations of degree m monomials in M_{ij} also form a vector space
- 1-Matrix monomials are symmetric

$$M_{i_1 j_1} M_{i_2 j_2} = M_{i_2 j_2} M_{i_1 j_1}$$

- Instead of $(V_D \otimes V_D)^{\otimes m}$ we have

$$\text{Sym}^m(V_D \otimes V_D) = \text{Symmetric part of } (V_D \otimes V_D)^{\otimes m}$$

- For example

$$M_{i_1 j_1} M_{i_2 j_2} \longleftrightarrow \left[\begin{aligned} &(|i_1\rangle \otimes |j_1\rangle) \otimes (|i_2\rangle \otimes |j_2\rangle) \\ &+ (|i_2\rangle \otimes |j_2\rangle) \otimes (|i_1\rangle \otimes |j_1\rangle) \end{aligned} \right]$$

1-Matrix Model

[Ramgoolam '18]

- Notation

$$W_{m,D} = \text{Sym}^m(V_D \otimes V_D)$$

- $W_{m,D}$ is also a vector space of bosonic oscillators
- For example: $W_{2,D}$

$$M_{i_1 j_1} M_{i_2 j_2} \longleftrightarrow \left[\begin{aligned} & (|i_1\rangle \otimes |j_1\rangle) \otimes (|i_2\rangle \otimes |j_2\rangle) \\ & + (|i_2\rangle \otimes |j_2\rangle) \otimes (|i_1\rangle \otimes |j_1\rangle) \end{aligned} \right] \longleftrightarrow M_{i_1 j_1}^\dagger M_{i_2 j_2}^\dagger |0\rangle$$

2-Matrix Model

[Barnes, AP, Ramgoolam '21]

- Matrix model with two D^2 matrices M, N .
- Zero-dimensional QFT

$$Z = \int dM dN e^{-\mathcal{S}(M,N)}.$$

- Free (Gaussian)

$$\mathcal{S}(M, N) \sim M + N + MM + NN + MN$$

- Permutation invariant

$$\mathcal{S}(M_{\sigma(i)\sigma(j)}, N_{\sigma(k)\sigma(l)}) = \mathcal{S}(M_{ij}, N_{kl}), \quad \sigma \in \mathcal{S}_D$$

2-Matrix Model

[Barnes, AP, Ramgoolam '21]

- Observables

$$f(M_{\sigma(i)\sigma(j)}, N_{\sigma(k)\sigma(l)}) = f(M_{ij}, N_{kl}),$$

- Expectation values from Wick's theorem

$$\langle f(M_{ij}, N_{kl}) \rangle = \frac{1}{Z} \int dM dN e^{-S(M,N)} f(M_{ij}, N_{kl}),$$

- Degree n monomials in N_{kl} form a vector space

$$\text{Sym}^n(V_D \otimes V_D)$$

- Mixed monomials are not symmetric in M and N

$$M_{i_1 j_1} N_{k_1 l_1} \neq M_{k_1 l_1} N_{i_1 j_1}$$

- Degree $m + n$ monomials form tensor product of symmetric parts

$$W_{m,n,D} = \text{Sym}^m(V_D \otimes V_D) \otimes \text{Sym}^n(V_D \otimes V_D)$$

- Example: $W_{2,2,D}$

$$M_{i_1 j_1} M_{i_2 j_2} N_{k_1 l_1} N_{k_2 l_2}$$



$$\left[(|i_1\rangle \otimes |j_1\rangle) \otimes (|i_2\rangle \otimes |j_2\rangle) + (|i_2\rangle \otimes |j_2\rangle) \otimes (|i_1\rangle \otimes |j_1\rangle) \right] \otimes \left[(|k_1\rangle \otimes |l_1\rangle) \otimes (|k_2\rangle \otimes |l_2\rangle) + (|k_2\rangle \otimes |l_2\rangle) \otimes (|k_1\rangle \otimes |l_1\rangle) \right]$$

- $W_{m,n,D}$ is also a vector space of bosonic oscillators

$$M_{i_1 j_1} \dots M_{i_m j_m} N_{k_1 l_1} \dots N_{k_n l_n} \longleftrightarrow M_{i_1 j_1}^\dagger \dots M_{i_m j_m}^\dagger N_{k_1 l_1}^\dagger \dots N_{k_n l_n}^\dagger |0\rangle$$

2-Matrix Observables

- Degree $m + n$ observables form a subspace of S_D -invariants

$$W_{m,n,D}^{V_0} = \{w \in W_{m,n,D} \mid \sigma w = w\}.$$

- The goal is to construct a basis for $W_{m,n,D}^{V_0}$
- For example, a basis for $W_{1,D}^{V_0}$ is

$$\sum_{i=1}^D |i\rangle \otimes |i\rangle \longleftrightarrow \sum_{i=1}^D M_{ii}, \quad \sum_{i,j=1}^D |i\rangle \otimes |j\rangle \longleftrightarrow \sum_{i,j=1}^D M_{ij}.$$

- A good start is compute the dimension of $W_{m,n,D}^{V_0}$.
- Use projector to compute dimension of subspace

$$P_{V_0}^{S_D} : W_{m,n,D} \rightarrow W_{m,n,D}^{V_0}$$

- Trace gives dimension

$$\dim(W_{m,n,D}^{V_0}) = \text{Tr}_{W_{m,n,D}} \left(P_{V_0}^{S_D} \right).$$

- Notation

$$\text{Dim}(m, n, D) \equiv \dim(W_{m,n,D}^{V_0})$$

$$\text{Dim}(m, D) \equiv \dim(W_{m,0,D}^{V_0})$$

- Average over group gives a projector to invariants

$$P_{V_0}^{S_D} = \frac{1}{|S_D|} \sum_{\sigma \in S_D} \sigma$$

- Check that it is invariant

$$\sigma' P_{V_0}^{S_D} = P_{V_0}^{S_D}.$$

- Compute trace

$$\text{Dim}(m, n, D) = \text{Tr}_{W_{m,n,D}} \left(P_{V_0}^{S_D} \right) =$$

$$\sum_{\substack{p \vdash D \\ q \vdash m \\ r \vdash n}} \frac{1}{\prod_{i=1}^D i^{p_i} p_i! \prod_{i=1}^m i^{q_i} q_i! \prod_{i=1}^n i^{r_i} r_i!} \prod_{i=1}^m \left(\sum_{l|i} l p_l \right)^{2q_i} \prod_{j=1}^n \left(\sum_{l|j} l p_l \right)^{2r_j}$$

m	n	Dim(m, n, D) for $D = 1, \dots, 2m + 2n + 5$
1	0	1, 2, 2, 2, 2, 2, 2
2	0	1, 6, 10, 11, 11, 11, 11, 11, 11
1	1	1, 8, 14, 15, 15, 15, 15, 15, 15
3	0	1, 10, 31, 47, 51, 52, 52, 52, 52, 52, 52
2	1	1, 20, 70, 107, 116, 117, 117, 117, 117, 117, 117, 117

When $D \geq 2m + 2n$, the dimension stabilizes

$$\text{Dim}(m, n, D \geq 2m + 2n,) = \text{Dim}(m, n, 2m + 2n)$$

1. What happens when $\text{Dim}(m, n, D)$ stabilizes?

- 1-matrix case [Kartsaklis, Ramgoolam, Sadrzadeh '17]

$$\text{Dim}(m, 0, 2m) = \begin{array}{l} \# \text{ directed graphs with } m \\ \text{edges on unlabeled vertices} \end{array}$$

- Basis for observables labeled by directed graphs
- 2-matrix case [Barnes, AP, Ramgoolam '21]

$$\text{Dim}(m, n, 2m + 2n) = \begin{array}{l} \# \text{ directed colored graphs with } m \\ \text{blue edges } n \text{ green edges on unlabeled} \\ \text{vertices} \end{array}$$

- Basis for observables labeled by directed colored graphs

2. What happens for $D \leq 2m + 2n$?

Observables from graphs

Each blue edge going from vertex i to vertex j is associated with a factor of M_{ij} . Similarly for green edges and N_{kl} . We then sum over the vertex labels from one to D .

For example

$$\sum_{i,j,k,l} M_{ij} M_{kl} \leftrightarrow \begin{array}{c} \bullet \quad \bullet \\ \curvearrowright \quad \curvearrowright \\ i \quad j \quad k \quad l \end{array}, \quad \sum_{i,j,k} M_{ij} N_{jk} \leftrightarrow \begin{array}{c} \bullet \quad \bullet \quad \bullet \\ \curvearrowright \quad \curvearrowright \\ i \quad j \quad k \end{array}$$

Proposition

$\text{Dim}(m, D)$ is the number of directed graphs with m edges on up to D unlabeled vertices.

Strategy

1. Start with labeled vertices
2. Unlabeled graphs correspond to equivalence classes of labeled graphs
3. Count equivalence classes
4. Relate counting to trace of projector

Step 1: Labeled graphs with m edges and k vertices

- Labeled graphs are collections of m ordered pairs of integers

$$[(a_1^-, a_1^+), \dots, (a_m^-, a_m^+)], \quad a_i^\pm \in \{1, \dots, k\}.$$

- Integers can take values in $\{1, \dots, k\}$
- Each pair is an edge from vertex a_i^- to vertex a_i^+ .
- Order of pairs in the collection does not matter.
- Order within pairs is important (directed edges).
- Example: $[(1, 2), (2, 1)]$



Step 2: Unlabeled graphs

- Two collections describe the same unlabeled graph if one is a re-labeling of the other.

Example

 $[(1, 2), (2, 1)]$ 

~

 $[(1, 3), (3, 1)]$  $[(1, 2), (2, 1)]$ 

≠

 $[(1, 2), (2, 3)]$ 

- Re-labeling is a S_k action on collections of pairs (labeled graphs)

$$\sigma^{-1} \cdot [(\mathbf{a}_1^-, \mathbf{a}_1^+), \dots, (\mathbf{a}_m^-, \mathbf{a}_m^+)] = [(\sigma(\mathbf{a}_1^-), \sigma(\mathbf{a}_1^+)), \dots, (\sigma(\mathbf{a}_m^-), \sigma(\mathbf{a}_m^+))].$$

Definition: Group orbit

If G acts on the set U , we say that $u \in U$ and $u' \in U$ are in the same orbit if there exists $g \in G$ such that $g \cdot u = u'$.

$$O_u = \{g \cdot u \mid \forall g \in G\}$$

- Group orbits partition the set U into disjoint subsets
- The set of all orbits is written as a quotient

$$G \backslash U$$

Example

Labeled graphs with 1 edge and 3 vertices under action of S_3 .

$$U_{1,3} = \{[(1, 1)], [(1, 2)], [(1, 3)], [(2, 1)], [(2, 2)], [(2, 3)], [(3, 1)], [(3, 2)], [(3, 3)]\}$$



$$O_{[(1,1)]} = \{[(1, 1)], [(2, 2)], [(3, 3)]\},$$

$$O_{[(1,2)]} = \{[(1, 2)], [(1, 3)], [(2, 1)], [(2, 3)], [(3, 1)], [(3, 2)]\}$$

Step 3: Count Orbits

- Let $U_{m,k}$ be the set of all labeled graphs with m edges and k vertices.
- Proposition is equivalent to the statement

$$\text{Dim}(m, k) = |S_k \setminus U_{m,k}|.$$

- Burnside's lemma: Number of orbits is equal to average number of fixed points

$$\begin{aligned} |S_k \setminus U_{m,k}| &= \frac{1}{|S_k|} \sum_{\sigma \in S_k} (\# \text{ Elements in } U_{m,k} \text{ fixed by } \sigma) \\ &= \frac{1}{|S_k|} \sum_{\sigma \in S_k} \sum_{u \in U_{m,k}} \delta(\sigma \cdot u = u) \end{aligned}$$

Example

$U_{2,2}$ consists of the 10 elements

$[(1, 1), (1, 1)]$ $[(1, 1), (1, 2)]$ $[(1, 1), (2, 1)]$ $[(1, 1), (2, 2)]$
 $[(1, 2), (1, 2)]$ $[(1, 2), (2, 1)]$ $[(1, 2), (2, 2)]$
 $[(2, 1), (2, 1)]$ $[(2, 1), (2, 2)]$
 $[(2, 2), (2, 2)]$

$S_2 = \{e, (12)\}$.

$$|S_2 \setminus U_{2,2}| = \frac{1}{2!}(10 + 2) = 6.$$

Step 4: Trace of projector

- $U_{m,k}$ defines permutation representation

$$V_{m,k} = \text{Span}(|u\rangle \mid u \in U_{m,k})$$

- S_k acts on $V_{m,k}$ by the action on $U_{m,k}$

$$\sigma |u\rangle = |\sigma \cdot u\rangle.$$

- It is the symmetric part of $(V_k \otimes V_k)^{\otimes m}$

$$V_{m,k} \cong \text{Sym}^m(V_k \otimes V_k)$$

- $V_{m,k}$ is a bosonic oscillator vector space

$$|u\rangle = |(a_1^-, a_1^+), \dots, (a_m^-, a_m^+)\rangle = A_{a_1^-, a_1^+}^\dagger \dots A_{a_m^-, a_m^+}^\dagger |0\rangle.$$

Step 4: Trace of projector

- Re-write the Kronecker delta

$$\delta(\sigma \cdot u = u) \longleftrightarrow \langle u | \sigma | u \rangle .$$

- Burnside's lemma becomes

$$\begin{aligned} |S_k \setminus U_{m,k}| &= \frac{1}{k!} \sum_{\sigma \in S_k} \sum_{u \in U_{m,k}} \langle u | \sigma | u \rangle \\ &= \sum_{u \in U_{m,k}} \langle u | P_{V_0}^{S_k} | u \rangle \\ &= \text{Tr}_{V_{m,k}}(P_{V_0}^{S_k}). \end{aligned}$$

- 2-colored directed graphs with m, n edges on k vertices

$$[(a_1^-, a_1^+), \dots, (a_m^-, a_m^+)], [(b_1^-, b_1^+), \dots, (b_n^-, b_n^+)],$$

$$a_i^\pm, b_i^\pm \in \{1, \dots, k\}$$

- $U_{m,n,k}$ defines permutation representation

$$V_{m,n,k} = \text{Span}(|u\rangle \mid u \in U_{m,n,k})$$

- $V_{m,n,k}$ is a bosonic oscillator space

$$|u\rangle = A_{a_1^- a_1^+}^\dagger \cdots A_{a_m^- a_m^+}^\dagger B_{b_1^- b_1^+}^\dagger \cdots B_{b_n^- b_n^+}^\dagger |0\rangle$$

- Number of orbits is equal to dimension of subspace of invariants

$$|S_k \setminus U_{m,n,k}| = \text{Tr}_{V_{m,n,k}}(P_{V_0}^{S_k})$$

- Replace k with D
- It is the same vector space of matrix polynomials

$$V_{m,D} \cong W_{m,D} = \text{Sym}^m(V_D \otimes V_D)$$

$$V_{m,n,D} \cong W_{m,n,D} = \text{Sym}^m(V_D \otimes V_D) \otimes \text{Sym}^n(V_D \otimes V_D)$$

- The action of S_D on W is the same as S_k on V

$$\text{Dim}(m, D) = \text{Tr}_{W_{m,D}}(P_{V_0}^{S_D}) = \text{Tr}_{V_{m,D}}(P_{V_0}^{S_D}) = |S_D \setminus U_{m,D}|$$

$$\text{Dim}(m, n, D) = \text{Tr}_{W_{m,n,D}}(P_{V_0}^{S_D}) = \text{Tr}_{V_{m,n,D}}(P_{V_0}^{S_D}) = |S_D \setminus U_{m,n,D}|.$$

- $\text{Dim}(m, D) = |S_D \setminus U_{m,D}|$ stabilizes for $D \geq 2m$, because the most number of vertices that can be occupied by m edges is $2m$.



- $\text{Dim}(m, n, D) = |S_D \setminus U_{m,n,D}|$ stabilizes because $m + n$ edges can occupy no more than $2m + 2n$ vertices

- We set out to understand a basis at finite D .
- For $D < 2m + 2n$, a basis is labeled by graphs with $m + n$ edges on D vertices.

$$\text{Dim}(m, n, D) = |S_D \setminus U_{m,n,D}|.$$

Summary: Observables

- Burnside's lemma using permutation representations

$$\text{Dim}(m, n, D) = |S_D \setminus U_{m,n,D}|$$

- Orbits are graphs with $m + n$ edges, D vertices
- Explicit bijection to observables

$$\sum_{i,j,k,l} M_{ij} M_{kl} \leftrightarrow \begin{array}{c} \overset{\curvearrowright}{\color{blue}i \quad j} \quad \overset{\curvearrowright}{\color{blue}k \quad l} \end{array}, \quad \sum_{i,j,k} M_{ij} N_{jk} \leftrightarrow \begin{array}{c} \overset{\curvearrowright}{\color{blue}i \quad j} \quad \overset{\curvearrowright}{\color{green}j \quad k} \end{array}$$

Expectation Values of Observables

When the VEV is non-zero, Wick's theorem takes the form

$$\langle T_{i_1 j_1} \dots T_{i_r j_r} \rangle = \sum_{p \in P_r^{1,2}} \prod_{c \in p} \prod_{(a,b) \in p} \langle T_{i_a j_a} T_{i_b j_b} \rangle_c \langle T_{i_c j_c} \rangle,$$

where T can be either M or N .

At degree $m = 2, n = 1$ we have

$$\begin{aligned}\langle M_{i_1 j_1} M_{i_2 j_2} N_{i_3 j_3} \rangle = & \langle M_{i_1 j_1} M_{i_2 j_2} \rangle_c \langle N_{i_3 j_3} \rangle + \langle M_{i_1 j_1} N_{i_3 j_3} \rangle_c \langle M_{i_2 j_2} \rangle \\ & + \langle M_{i_2 j_2} N_{i_3 j_3} \rangle_c \langle M_{i_1 j_1} \rangle + \langle M_{i_1 j_1} \rangle \langle M_{i_2 j_2} \rangle \langle N_{i_3 j_3} \rangle\end{aligned}$$

- $V_D \otimes V_D$ decomposes into irreducible representations of S_D

$$V_D \otimes V_D \longrightarrow 2V_0 \oplus 3V_H \oplus V_2 \oplus V_3$$

- Use Clebsch-Gordan coefficients

$$S_a^{V_A; \alpha} = \sum_{i,j} C_{a,ij}^{V_A; \alpha} M_{ij}, \quad a = 1, \dots, \dim(V_A)$$

- They transform irreducibly under S_D

$$\sum_{i,j} C_{a,ij}^{V_A; \alpha} M_{\sigma^{-1}(i)\sigma^{-1}(j)} = D^{V_A}(\sigma)_a^b S_b^{V_A; \alpha}.$$

- Correlators are simple in the representation basis

$$\langle S_a^{V_A;\alpha} S_b^{V_B;\beta} \rangle_c = \left(\Lambda_A^{MM} \right)_{\alpha\beta}^{-1} \delta_{ab} \delta(V_A, V_B).$$

- We have symmetric coupling matrices

$$\frac{2 \cdot 3}{2} + \frac{3 \cdot 4}{2} + 1 + 1 = 11$$

- Invert the change of basis

$$M_{ij} = \sum_{V_A, a, \alpha} C_{a,ij}^{V_A; \alpha} S_a^{V_A; \alpha}$$

- The two point function becomes

$$\begin{aligned} \langle M_{ij} M_{kl} \rangle_c &= \sum_{V_A, a, \alpha} \sum_{V_B, b, \beta} C_{a,ij}^{V_A; \alpha} \langle S_a^{V_A; \alpha} S_b^{V_B; \beta} \rangle_c C_{b,kl}^{V_B; \beta} \\ &= \sum_{V_A, a, \alpha, \beta} C_{a,ij}^{V_A; \alpha} \left(\Lambda_A^{MM} \right)_{\alpha\beta}^{-1} C_{a,kl}^{V_A; \beta} \end{aligned}$$

- Repeat for N

$$R_a^{V_A;\alpha} = \sum_{i,j} C_{a,ij}^{V_A;\alpha} N_{ij}$$

- The correlators are simple in this basis

$$\langle R_a^{V_A;\alpha} R_b^{V_B;\beta} \rangle_c = \left(\Lambda_A^{NN} \right)_{\alpha\beta}^{-1} \delta_{ab} \delta(V_A, V_B),$$

$$\langle S_a^{V_A;\alpha} R_b^{V_B;\beta} \rangle_c = \left(\Lambda_A^{MN} \right)_{\alpha\beta}^{-1} \delta_{ab} \delta(V_A, V_B).$$

- $\left(\Lambda_A^{MN} \right)_{\alpha\beta}$ is not a symmetric matrix.

$$2^2 + 3^2 + 1 + 1 = 15$$

- Inverting the change of basis

$$\langle N_{ij} N_{kl} \rangle_c = \sum_{V_A, a, \alpha, \beta} C_{a,ij}^{V_A;\alpha} \left(\Lambda_A^{NN} \right)_{\alpha\beta}^{-1} C_{a,kl}^{V_A;\beta}$$

$$\langle M_{ij} N_{kl} \rangle_c = \sum_{V_A, a, \alpha, \beta} C_{a,ij}^{V_A;\alpha} \left(\Lambda_A^{MN} \right)_{\alpha\beta}^{-1} C_{a,kl}^{V_A;\beta}$$

- In general

$$\langle T_{ij} T_{kl} \rangle_c = \sum_{V_A, a, \alpha, \beta} C_{a,ij}^{V_A; \alpha} \left(\Lambda_A^{TT} \right)_{\alpha\beta}^{-1} C_{a,kl}^{V_A; \beta}$$

- We need to compute the tensors

$$\sum_a C_{a,ij}^{V_A; \alpha} C_{a,kl}^{V_A; \beta}$$

- They can all be written in terms of the same tensor

$$F_{ij} = \delta_{ij} - \frac{1}{D}$$

- Also true about VEV

- We had an orthonormal basis for $V_0 \oplus V_H$

$$|E_0\rangle = \sum_{i=1}^D C_{0i} |i\rangle = \frac{1}{\sqrt{D}} \sum_{i=1}^D |i\rangle$$

$$|E_a\rangle = \sum_{i=1}^D C_{ai} |i\rangle, \quad a = 1, \dots, D-1$$

- F_{ij} are matrix elements of P_{V_H}

$$F_{ij} = \sum_{a=1}^{D-1} \langle i|E_a\rangle \langle E_a|j\rangle = \delta_{ij} - \frac{1}{D}.$$

- Derived from

$$\langle i|j\rangle = \delta_{ij} = \langle i| \left(|E_0\rangle \langle E_0| + \sum_{a=1}^{D-1} |E_a\rangle \langle E_a| \right) |j\rangle$$

- There are 15 combinations, here are a select few

$$\sum_a C_{a,ij}^{V_0;2} C_{a,kl}^{V_0;2} = \frac{F_{ij} F_{kl}}{D-1}$$

$$\sum_a C_{a,ij}^{V_H;1} C_{a,kl}^{V_H;1} = \frac{F_{jl}}{D}$$

$$\sum_a C_{a,ij}^{V_H;3} C_{a,kl}^{V_H;3} = \sum_{p,q} F_{ip} F_{jp} F_{pq} F_{qk} F_{ql}$$

$$\sum_a C_{a,ij}^{V_3;1} C_{a,kl}^{V_3;1} = \frac{1}{2} (F_{ik} F_{jl} - F_{il} F_{jk})$$

- Expectation values are weighted sums of products of F
- Some contributions to $\langle M_{i_1 j_1} M_{i_2 j_2} N_{i_3 j_3} \rangle$ are proportional to

$$F_{i_1 j_1} F_{i_2 j_2} F_{i_3 j_3},$$
$$\sum_{p,q=1}^D F_{i_1 p} F_{j_1, p} F_{p q} F_{i_2 q} F_{j_2 q} F_{i_3 j_3}.$$

- Observables have all external indices summed.

$$\langle \text{●} \begin{array}{c} \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \\ \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \\ \text{---} \text{---} \text{---} \text{---} \text{---} \text{---} \end{array} \text{●} \rangle = \sum_{i,j} \langle M_{ij} M_{ij} N_{ij} \rangle$$

- The F contributions become

$$\sum_{i,j=1}^D F_{ij} F_{ij} F_{ij},$$
$$\sum_{i,j,p,q=1}^D F_{ip} F_{jp} F_{pq} F_{iq} F_{jq} F_{ij}.$$

The general strategy is to decompose the sums into parts

$$\sum_{i,j=1}^D = \sum_{i=j} + \sum_{i \neq j},$$

$$\sum_{i,j,k=1}^D = \sum_{i=j=k} + \sum_{i=j \neq k} + \sum_{i=k \neq j} + \sum_{j=k \neq i} + \sum_{i \neq j \neq k}.$$

Because F_{ij} is almost like a Kronecker delta,

$$F_{ij} = \begin{cases} 1 - \frac{1}{D} & i = j \\ -\frac{1}{D} & i \neq j \end{cases}$$

it is constant on each part

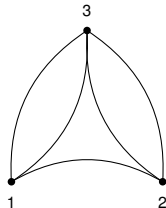
Example 1

$$\begin{aligned}\sum_{i,j} (F_{ij})^A &= \sum_{i=j} (F_{ij})^A + \sum_{i \neq j} (F_{ij})^A \\ &= D \left[1 - \frac{1}{D} \right]^A + D(D-1) \left[\frac{1}{(-D)} \right]^A \\ &= \left[D(1-D)^A + D(D-1) \right] \frac{1}{(-D)^A}.\end{aligned}$$

Example 2

$$\begin{aligned}
 \sum_{i_1, i_2, i_3=1}^D F_{i_1 i_2}^{A_{12}} F_{i_1 i_3}^{A_{13}} F_{i_2 i_3}^{A_{23}} &= \left(\sum_{123} + \sum_{12|3} + \sum_{13|2} + \sum_{23|1} + \sum_{1|2|3} \right) F_{i_1 i_2}^{A_{12}} F_{i_1 i_3}^{A_{13}} F_{i_2 i_3}^{A_{23}} \\
 &= \frac{D!}{(D-1)!} \left[1 - \frac{1}{D} \right]^{A_{12}+A_{13}+A_{23}} + \frac{D!}{(D-2)!} \left[1 - \frac{1}{D} \right]^{A_{12}} \left[\frac{1}{-D} \right]^{A_{13}+A_{23}} \\
 &+ \frac{D!}{(D-2)!} \left[1 - \frac{1}{D} \right]^{A_{13}} \left[\frac{1}{-D} \right]^{A_{12}+A_{23}} + \frac{D!}{(D-2)!} \left[1 - \frac{1}{D} \right]^{A_{23}} \left[\frac{1}{-D} \right]^{A_{12}+A_{13}} \\
 &+ \frac{D!}{(D-3)!} \left[\frac{1}{-D} \right]^{A_{12}+A_{13}+A_{23}} \\
 &= \left[\frac{1}{-D} \right]^{A_{12}+A_{13}+A_{23}} \left(\frac{D!(1-D)^{A_{12}+A_{13}+A_{23}}}{(D-1)!} \right. \\
 &+ \frac{D!(1-D)^{A_{12}}}{(D-2)!} + \frac{D!(1-D)^{A_{13}}}{(D-2)!} + \frac{D!(1-D)^{A_{23}}}{(D-2)!} + \left. \frac{D!}{(D-3)!} \right)
 \end{aligned}$$

- The procedure can be organized in terms of undirected graphs
- For example,

$$\sum_{i_1, i_2, i_3=1}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2 \longleftrightarrow$$


- Sum over p indices gives graph on p vertices
- Each factor F_{i_a, i_b} is an edge from a to b

- General products with p indices summed

$$\sum_{i_1, \dots, i_p} \prod_{1 \leq a < b \leq p} F_{i_a i_b}^{A_{ab}}$$

- Parameterized by matrix A_{ab}
- Interpret A_{ab} as an adjacency matrix

- Decomposition

$$\sum_{i_1, i_2, i_3=1}^D = \sum_{i_1=i_2=i_3} + \sum_{i_1=i_2 \neq i_3} + \sum_{i_1=i_3 \neq i_2} + \sum_{i_2=i_3 \neq i_1} + \sum_{i_1 \neq i_2 \neq i_3}$$

- Corresponds to set partitions

$$i_1 = i_2 = i_3 \leftrightarrow \{123\}, \quad i_1 = i_2 \neq i_3 \leftrightarrow \{12\}, \{3\},$$

$$i_1 \neq i_2 \neq i_3 \leftrightarrow \{1\}, \{2\}, \{3\}$$

- Notation for sums

$$\sum_{i_1, i_2, i_3=1}^D = \sum_{\{123\}} + \sum_{\{12\}, \{3\}} + \sum_{\{13\}, \{2\}} + \sum_{\{23\}, \{1\}} + \sum_{\{1\}, \{2\}, \{3\}}$$

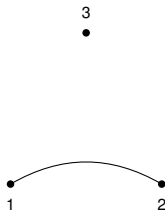
- Number of terms in sum with b blocks is

$$\frac{D!}{(D-b)!}$$

- Compute each sum using graphs operations

$$\sum_{i_1=i_2 \neq i_3} F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2 \leftrightarrow \sum_{\{12\}, \{3\}}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2$$

- Sum is associated with a new graph

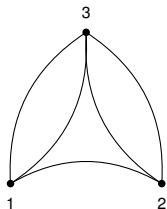


$$= \frac{D!}{(D-2)!} (1-D)^1 \frac{1}{(-D)^5}$$

- All indices equal

$$\sum_{i_1=i_2=i_3}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2 \leftrightarrow \sum_{\{123\}}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2$$

- Associated graph contributes


$$= \frac{D!}{(D-1)!} (1-D)^5 \frac{1}{(-D)^5}$$

- i_1 and i_3 equal

$$\sum_{i_1=i_3 \neq i_2}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2 \leftrightarrow \sum_{\{13\}, \{2\}}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2$$

- Associated graph contributes

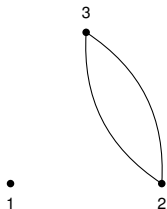


$$= \frac{D!}{(D-2)!} (1-D)^2 \frac{1}{(-D)^5}$$

- i_2 and i_3 equal

$$\sum_{i_2=i_3 \neq i_1}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2 \leftrightarrow \sum_{\{23\}, \{1\}}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2$$

- Associated graph contributes



$$= \frac{D!}{(D-2)!} (1-D)^2 \frac{1}{(-D)^5}$$

- All different

$$\sum_{i_1 \neq i_2 \neq i_3}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2 \leftrightarrow \sum_{\{1\}, \{2\}, \{3\}}^D F_{i_1 i_2}^1 F_{i_1 i_3}^2 F_{i_2 i_3}^2$$

- Associated graph contributes

$$\begin{array}{c} 3 \\ \bullet \\ \\ \\ \\ \\ \\ \\ \\ 1 \quad \quad 2 \end{array} = \frac{D!}{(D-3)!} (1-D)^0 \frac{1}{(-D)^5}$$

Summary: Expectation Values

- Expectation values are computed using Wick's theorem
- They ultimately boil down to products of F with all indices summed
- We organized the sums using undirected graphs
- Procedure manifestly independent of D

Summary

- We proved that observables correspond to directed colored graphs
- The one-point and two-point function are simple in a representation basis
- Computing expectation values is largely about the interplay between the graph and representation basis
- The interplay can be described using undirected uncolored graphs.

Outlook

- Theory
 - Extend the permutation invariant program to tensor models
 - Correlators in the matrix models
 - Interesting parameter limits
 - Double cosets
- Applied
 - Apply 2-matrix model to data
 - Improve algorithm

Outlook

Thank you for listening!