

# A Holant Dichotomy: Is the FKT Algorithm Universal?

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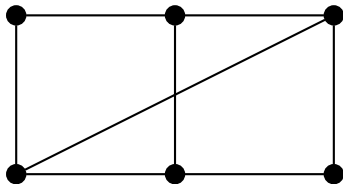
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Oct 20, 2015

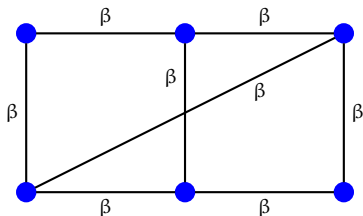
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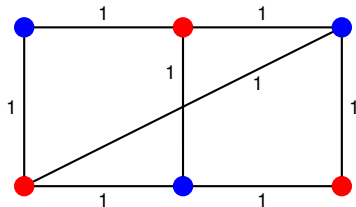
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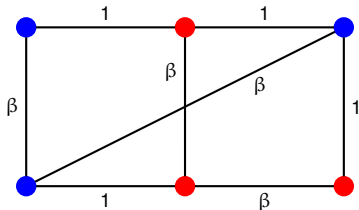
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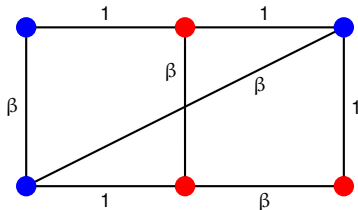
Configuration  $\sigma : V \rightarrow \{0, 1\}$

$$w(\sigma) = \beta^4$$

$$\Pr(\sigma) \sim w(\sigma)$$

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Partition function (normalizing factor):

$$Z_G(\beta) = \sum_{\sigma: V \rightarrow \{0,1\}} w(\sigma)$$

where  $w(\sigma) = \beta^{m(\sigma)}$ ,  $m(\sigma)$  is the number of monochromatic edges under  $\sigma$ .

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- For **planar** graphs, there is a polynomial time algorithm [[Kastelyn 61 & 67](#), [Temperley and Fisher 61](#)].
- Reduction to **#PM** (counting perfect matchings) in planar graphs.
  - ▶ **#PM** is **#P**-hard [[Valiant 79](#)] in **general** graphs as well.
- **#PM** can be computed via Pfaffian orientations of planar graphs.

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We need to answer this question in some framework.

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- **Name**  $\#CSP(\mathcal{F})$

**Instance** A bipartite graph  $G = (V, C, E)$  and a mapping  $\pi : C \rightarrow \mathcal{F}$

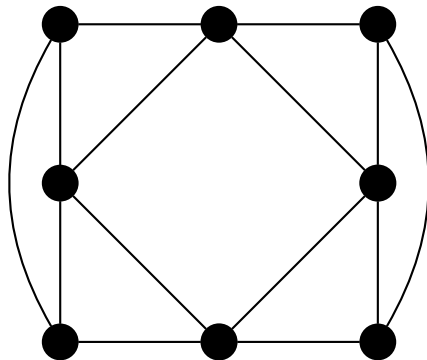
**Output** The quantity:

$$\sum_{\sigma: V \rightarrow \{0,1\}} \prod_{c \in C} f_c(\sigma|_{N(c)}),$$

where  $N(c)$  are the neighbors of  $c$  and  $f_c = \pi(c) \in \mathcal{F}$ .

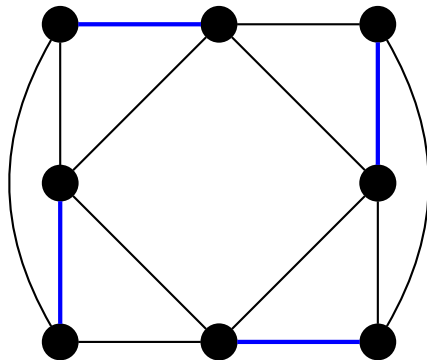
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- Edge-coloring models — edges are variables and vertices are functions.
- **Name** Holant( $\mathcal{F}$ )

**Instance** A graph  $G = (V, E)$  and a mapping  $\pi : V \rightarrow \mathcal{F}$

**Output** The quantity:

$$\sum_{\sigma: E \rightarrow \{0,1\}} \prod_{v \in V} f_v(\sigma|_{E(v)}),$$

where  $E(v)$  are the incident edges of  $v$  and  $f_v = \pi(v) \in \mathcal{F}$ .

- More general than #CSP:

$$\#CSP(\mathcal{F}) \equiv_T \text{Holant}(\mathcal{EQ} \cup \mathcal{F}),$$

where  $\mathcal{EQ} = \{=_1, =_2, =_3, \dots\}$  is the set of equalities of all arities.

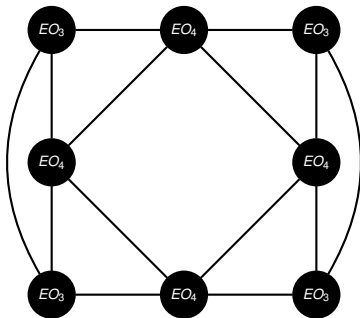
- Equivalent formulation: Tensor network contraction ...
- PI-Holant( $\mathcal{F}$ ) denotes the version where instances are all **planar**.

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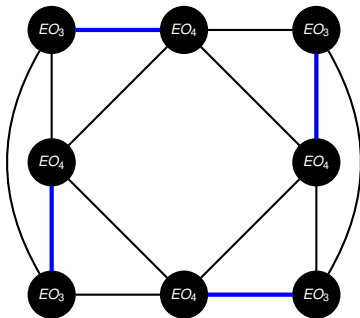
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## #PM as a Holant

- Put functions EXACTONE (EO) on nodes (edges are variables).
- #PM is then the partition function:

$$\#PM = \sum_{\sigma: E \rightarrow \{0,1\}} \prod_{v \in V} \text{EO}_d(\sigma|_{E(v)}).$$



## Complexity Classifications

Counting problems with local constraints are usually classified into:

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Category (2) is always captured by holographic algorithms with matchgates.

Examples include:

- Tutte polynomials [Vertigan 91], [Vertigan 05].
- Spin systems [Kowalczyk 10], [Cai, Kowalczyk, Williams 12].
- **#CSP** [Cai, Lu, Xia 10], [G. and Williams 13].



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Category (1) is characterized in [Cai, G., Williams 13].

Category (3) is **not** captured by holographic algorithms with matchgates!

## New Planar Tractable Case

Counting Orientations, where two types of nodes are allowed:

1. Exactly one edge coming in;
2. All edges coming in or going out (either a sink or a source).

Moreover, we require that the **gcd** of the degrees of **type 2** nodes is at least **5**.

Then the problem is tractable.

## #PM in Planar Hypergraphs

As a special case of our result, consider the following problem.

**Name** #Planar-Hyper-PM( $S$ )

**Instance** A hypergraph  $H$  whose incidence graph is planar, and hyperedge sizes are prescribed by  $S$ .

**Output** The number of perfect matchings in  $H$ .



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Let  $t = \gcd(S)$ .

- If  $t \geq 5$  or  $S \subseteq \{1, 2\}$ , then #Planar-Hyper-PM( $S$ ) is computable in polynomial time.
- Otherwise  $t \leq 4$ ,  $S \not\subseteq \{1, 2\}$ , and #Planar-Hyper-PM( $S$ ) is #P-hard.

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- Some steps of the process may provide orientations **inconsistent** with the original instance, but we can keep track of enough information to go back and check.
- Tractable mainly due to degree rigidity.

# Exemplary Planar Structures

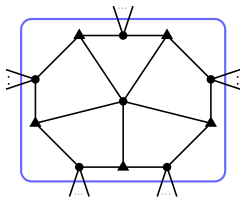
## Lemma

Let  $G = (L \cup R, E)$  be a planar bipartite graph with parts  $L$  and  $R$ .

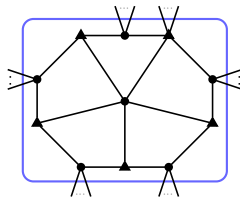
Every vertex in  $L$  has degree at least 5;

every vertex in  $R$  has degree at least 3.

If  $G$  is simple, then there exists one of the two wheel structures in  $G$ .



(a) Type 1



(b) Type 2

# A Score Based Proof

- Assign a *score*  $s_v$  to each vertex  $v \in V$  so that

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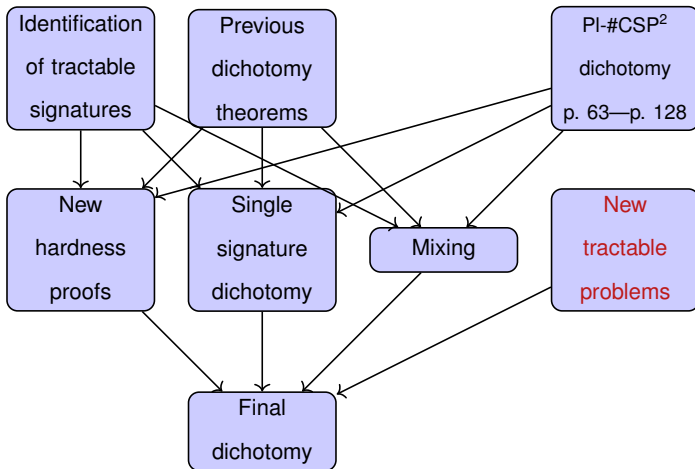
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- If no wheel structure exists, then there exists a 1-1 mapping between positive vertices and negative vertices, and negative scores are larger. Hence the total score has to be negative. Contradiction.

# Proof Roadmap of the Main Theorem



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- The natural generalization for  $d \geq 5$  does not hold, and in the end we proved the  $d = 2$  case (where the natural generalization does hold).
- However lots of progress was made due to this belief.

## Concluding Remarks

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- There exists planar tractable cases that are **not** captured by holographic algorithms with matchages (or FKT).



Thank You!