A very sharp threshold for first order logic distinguishability of random graphs

Itai Benjamini, Maksim Zhukovskii
Weizmann Institute of Science

Abstract

In this paper we find an integer $h = h(n)$ such that the minimum number of variables of a first order sentence that distinguishes between two independent uniformly distributed random graphs of size $n$ with the asymptotically largest possible probability $\frac{1}{4} - o(1)$ belongs to $\{h, h+1, h+2, h+3\}$. We also prove that the minimum (random) $k$ such that two independent random graphs are distinguishable by a first order sentence with $k$ variables belongs to $\{h, h+1, h+2\}$ with probability $1 - o(1)$.

1 Introduction

We say that $G_1, G_2$ are $k$-distinguishable, if there exists a first order (FO) sentence $\varphi$ with $k$ variables such that $G_1 \models \varphi$ while $G_2 \not\models \varphi$. In this paper we address the following question. Let $G_1^n, G_2^n$ be chosen uniformly at random and independently from the set of all graphs on $[n] := \{1, \ldots, n\}$. What is the minimum $k$ such that they are $k$-distinguishable?

The classical FO 0-1 law [5, 7] says that, for every fixed sentence, its truth value on the random graph on $n$ vertices approaches either 0 or 1 as $n \to \infty$. Therefore, fixed FO sentences do not distinguish between random graphs with high probability. However, if we allow the number of variables to grow with $n$, then this is not true. In this paper, we find the minimum number of variables of a FO sentence that distinguishes between two independent random graphs (up to a constant additive term bounded by 4).

Note that the number of variables is an important measure of complexity of a FO sentence. In particular, the problem of deciding whether a graph on $n$ vertices satisfies a FO sentence with $k$ variables can be solved in time $O(n^k)$ (see [12 Proposition 6.6]).

If we fix a sequence of sentences $\varphi = \varphi(n)$ in advance, then due to the independency, $P(G_1^n \models \varphi, G_2^n \not\models \varphi) \leq \frac{1}{4}$. However, if we choose $\varphi$ depending on the graphs (i.e. $\varphi$ is random), then it may happen that the graphs are distinguishable with probability close to 1. Below we give very sharp answers to the following questions.
1. What is the minimum (not random) number of variables of a FO sentence \( \varphi \) such that 
\[ \Pr(G_1^n \models \varphi, G_2^n \not\models \varphi) \] is bounded away from 0 (or even equals \( 1/4 - o(1) \))?

2. What is the minimum (random) \( k \) such that \( G_1^n, G_2^n \) are \( k \)-distinguishable?

To get a naive lower bound in both questions (that turns out to be very sharp), let us recall that the graphs satisfying \( k \)-extension axioms (that we define below) are not \((k + 1)\)-distinguishable. The extension axiom \( \varphi_k \) is the FO sentence with \( k + 1 \) variables saying that, for every \( k \)-set and its partition \( A \cup B \), there exists a vertex \( x \) outside the set adjacent to all vertices in \( A \) and no vertex from \( B \). Due to the following claim (see \([12, \text{Lemma 12.7}]\)), in order to prove that whp \( G_1^n, G_2^n \) are not \( k \)-distinguishable, it is sufficient to prove that with probability \( 1 - o(1) \) (with high probability or briefly whp) the uniformly sampled random graph \( G_n \) on \([n]\) satisfies \( \varphi_{k-1} \).

**Claim 1** If \( G_1, G_2 \) satisfy \( \varphi_k \), then there is no sentence with \( k + 1 \) variables distinguishing between \( G_1, G_2 \).

By the union bound
\[
\Pr(G_n \not\models \varphi_k) \leq \binom{n}{k} 2^k \left( 1 - \frac{1}{2^k} \right)^{n-k} = e^{f^{\text{ext}}(k)},
\]
where
\[
f^{\text{ext}}(k) = k \ln n - k \ln k + k \ln 2 - n2^{-k} + O(\ln k).
\]
Note that \( f^{\text{ext}}(k) > k(1 + o(1)) \) if \( k > \log_2 n \) and, for all \( k \),
\[
f^{\text{ext}}(k + 1) - f^{\text{ext}}(k) = \ln n + n2^{-k-1} - O(\ln k).
\]
Therefore, there is a unique \( \hat{k} \in [1, n] \) such that \( f^{\text{ext}}(\hat{k}) = 0 \):
\[
\hat{k} = \log_2 n - 2 \log_2 \ln n + \log_2 \ln 2 + o(1).
\]
Clearly, \( e^{f^{\text{ext}}(\lfloor \hat{k} \rfloor)} \leq 1 + o(1) \) and \( e^{f^{\text{ext}}(\lfloor \hat{k} \rfloor - 1)} = o(1) \). From this, we immediately get that whp \( G_1^n, G_2^n \) are not \( \lfloor \hat{k} \rfloor \)-distinguishable. It turns out that the lower bound is sharp up to a constant additive term.

**Theorem 1** Let \( \hat{k} \in [1, n] \) be the unique root of the equation \( f^{\text{ext}}(k) = 0 \).

1. The minimum \( k \) such that there exists a FO sentence \( \varphi \) with \( k \) variables satisfying \( \Pr(G_1^n \models \varphi, G_2^n \not\models \varphi) = \frac{1}{4} - o(1) \) belongs to \( \{\lfloor \hat{k} \rfloor + 1, \ldots, \lfloor \hat{k} \rfloor + 4\} \).

2. The minimum \( k \) such that \( G_1^n \) and \( G_2^n \) are \( k \)-distinguishable belongs to \( \{\lfloor \hat{k} \rfloor + 1, \lfloor \hat{k} \rfloor + 2, \lfloor \hat{k} \rfloor + 3\} \) whp.
The lower bound in both parts of Theorem 1 is already proven. Note that a weaker but asymptotically sharp upper bound \( \log_2 n(1+o(1)) \) (it differs from the lower bound by \( \Theta(\ln \ln n) \)) in the second part of Theorem 1 follows from the upper bound for the minimum quantifier depth of a FO sentence that defines a random graph whp [11]. The reason of why the random graph whp can be defined by a FO sentence with quantifier depth \( O(\log n) \) is that whp every two vertices have non-isomorphic neighborhoods, and so the upper bound is obtained by recursion, where for every vertex we write a sentence describing its neighborhood. Our proof strategy is absolutely different. We prove the first part of Theorem 1 in Section 2. In Section 3 we show that the same approach as in the proof in Section 2 can be used to prove part 2 of Theorem 1.

2 Proof of Theorem 1 part 1

The desired sentence that distinguishes between two random graphs says that there exists a set \( S \) that induces a specific subgraph with some additional restrictions on adjacencies between this set and all the other vertices. Let us first note that we are forced to make this property monotone (by the reason explained further). For that, we make the restrictions on subgraphs that induced by \( S \) to be downward closed. In other words, we construct a sequence of families of graphs \( \mathcal{F}_0 \subset \mathcal{F}_1 \subset \ldots \) on the vertex set \([k]\) such that \( \mathcal{F}_i \) is downward closed and also closed under isomorphism. It can be easily done by adding at each step a full single isomorphism class such that its number of edges is the same as on the previous step if possible, and has one more edge otherwise (the initial family is just a singleton consisting of an empty graph). For the random variable \( X(i) \) that counts the number of sets \( S \subset [n] \) such that \( G_n[S] \) (subgraph of \( G_n \) induced on \( S \)) is isomorphic to a graph from \( \mathcal{F}_i \), and there is no vertex \( x \) outside \( S \) adjacent to all vertices in \( S \). Note that for convenience and in order to making the considered events monotone, here we weaken the ‘extendibility’ condition 3

2.1 Expectation

Let \( k \) be large and let \( \mathcal{F}_k \) be a family of graphs on \([k]\) closed under isomorphism. Consider a uniformly distributed random graph \( G_n \) on \([n]\), and let \( X^\mathcal{F}_k \) be the total number of \( k \)-sets \( S \subset [n] \) such that \( G_n[S] \) (subgraph of \( G_n \) induced on \( S \)) is isomorphic to a graph from \( \mathcal{F}_k \), and there is no vertex \( x \) outside \( S \) adjacent to all vertices in \( S \). Note that for convenience and in order to making the considered events monotone, here we weaken the ‘extendibility’ condition
in the extension axioms (we consider only the partition \( S = A \sqcup B \), where \( A = S \), and \( B = \emptyset \)). Luckily it does not affect significantly the upper bound. We get

\[
E X^F_{[k]} = \binom{n}{k} \frac{|F_k|}{2^{k(k-1)/2}} \left( 1 - \frac{1}{2^k} \right)^{n-k} = e^{f_F(k)},
\]

\[
f_F(k) = k \ln n - k \ln k + k + \ln |F_k| - \frac{k(k-1)}{2} \ln 2 - n 2^{-k} + O(\ln k).
\]

Let \( k_F \in \mathbb{R} \) be such that \( f_F(k_F) = 0 \).

Let \( F \) be a graph on \([k]\) with a trivial automorphism group (it is easy to see that such a graph exists for large enough \( k \) — in particular, the random graph has no symmetries whp \([4]\)) and \( F_k^* \) contains all isomorphic copies of \( F \), i.e. \(|F_k^*| = k!\). Let \( F^0_k \subset F^1_k \subset \ldots \) be a sequence of graph families on \([k]\) considered in the beginning of the proof, i.e.

1. \( F^0_k \) is a singleton set containing only an empty graph,
2. for every \( i \), \( F^i_k \setminus F^{i-1}_k \) is a set of all isomorphic copies of a single graph \( F_i \) on \([k]\),
3. for every \( i \), \( F^i_k \) is downward closed (i.e., if \( H_1 \subset H_2 \) are graphs on \([k]\), and \( H_2 \in F^i_k \), then \( H_1 \in F^i_k \) as well).

Note that, for every \( i \), \(|F^i_k| - |F^{i-1}_k| \leq k!\) implying that \( E X_{F^i_k} - E X_{F^{i-1}_k} \leq E X_{F^*} \). If we can prove that, for some specific \( k \), \( E X^i_k \to 0 \) while, for large enough \( i \), \( E X^i_k \to \infty \), then we may hope that there exists \( i \) such that \( E X^i_k \to \ln 2 \). To show this, it is sufficient to prove that \( E X^i_k \to 0 \).

Now we let \( F_k = F_k^* \) and \( k = k^* \). Recall that \( E X^F_{[k]} \leq 1 \) and \( E X^F_{[k]} \geq 1 \). Let \( \Delta := k - [k] \).

Then

\[
f_F(k) - f_F([k]) = k^2 \ln 2 \left( 2^\Delta - 1 + o(1) \right).
\]

In order to be certain that \( E X^F_{[k]} \to 0 \), we need \( 2^\Delta - 1 \) to be bounded away from 0. Moreover, in order to have \( E X^F_{[k]} \to \infty \) for some \( i \), we need \( 2^\Delta - 1 \) not to be very large. The sufficient restriction \( 0.05 < \Delta \leq 0.55 \) is considered in the next section. For other values of \( \Delta \), we might slightly modify the sentence (and this is the reason why we get concentration in 4 points instead of 2).

\subsection*{2.1.1} \( 0.05 < \Delta \leq 0.55 \)

If \( \Delta > 0.5 \), then \( E X^F_{[k]} = o(1) \). If, at the same time, \( \Delta \leq 0.55 \), then \( 2^\Delta - 1 < \frac{1}{2} \). Therefore,

\[
\frac{2^{(|[k]|)}}{k!} E X^F_{[k]} \gg e^{f_F(k)} = 1
\]

implying that \( E X^F_{[k]} \to \infty \), where \( F^*_k \) is the set of all graphs on \([k]\). Let us recall that, for every \( i \),

\[
E X^i_{[k]} + E X^0_{[k]} \leq E X^i_{[k]} \leq E X^i_{[k]} + E X^i_{[k]}.
\]
Note also that $E[X_{[k]}^{F^0}] = \frac{1}{k} E[X_{[k]}^{F^*}]$ implying that $E[X_{[k]}^{F^0}] = o(1)$. Due to (2), we get that there exists $\mu = \mu(k)$ such that

$$E[X_{[k]}^{F^\mu}] = \ln 2 + o(1)$$

as needed. Note that $\{X_{[k]}^{F^\mu} = 0\}$ is expressed by a FO sentence with $|k| + 1 = |k - 0.05| + 1 \leq |\hat{k}| + 2$ variables. The latter inequality follows from the fact that $f_{F^0}(\hat{k} + 1.05) \rightarrow \infty$ implying that $k < \hat{k} + 1.05$. We compute the second moment of $X_{[k]}^{F^\mu}$ in Section 2.2.2.

### 2.1.2 $0.55 < \Delta \leq 0.95$

If $0.55 < \Delta \leq 0.95$, then our goal is to modify the definition of $X_k$ in a way such that the factor $(1 - \frac{1}{2k})^{-k}$ becomes $(1 - \frac{3}{2k})^{-k}$. Having that, we will get that $k_\Delta$ becomes roughly $k_\Delta + \log_2 3$, and $\Delta$ becomes roughly $\Delta + \log_2 3 - 2 \leq \log_2 3 - 1.05 < 0.55$.

Let $Y_k^{\mathcal{F}}$ be the total number of tuples $(v_1, v_2, S)$, where $v_1 \neq v_2 \in [n], S \in ([n]\{v_1, v_2\})$ such that $G_{n}[S]$ is equal to a graph from $\mathcal{F}_k$, and there are no vertices $x$ outside $S \cup \{v_1, v_2\}$ adjacent to all vertices from $S$ and at least 1 vertex among $v_1, v_2$. We have

$$EY_k^{\mathcal{F}} = \left(\frac{n}{2}\right)\binom{n-2}{k} |\mathcal{F}_k| \left(1 - \frac{3}{2^{k+2}}\right)^{n-k-2} = e^{g_\mathcal{F}(k)},$$

$$g_\mathcal{F}(k) = (k + 2) \ln n - k \ln k + k + \ln |\mathcal{F}_k| - \frac{k(k-1)}{2} \ln 2 - 3n2^{-k-2} + O(\ln k).$$

Let $k_\mathcal{F}' \in \mathbb{R}$ be such that $g_\mathcal{F}(k_\mathcal{F}') = 0$. Note that

$$k_\mathcal{F}' = k_\mathcal{F} + \log_2 \frac{3}{4} + o(1) = k_\mathcal{F} + \log_2 3 - 2 + o(1).$$

As above, letting $\mathcal{F}_k = \mathcal{F}_k^*$ and $k' = k_\mathcal{F}'$, we get that

$$\Delta' := k' - |k'| = \Delta + \log_2 \frac{3}{2} + o(1) - \left[\Delta + \log_2 \frac{3}{2} + o(1)\right] \in (0.1, 0.55)$$

since $0.55 < \Delta \leq 0.95$. Then $EY_{[k']}^{\mathcal{F}} = o(1)$ and

$$g_\mathcal{F}(k') - g_\mathcal{F}(|k'|) = (k')^2 \ln 2 \left(2^{\Delta'} - 1 + o(1)\right) < (k')^2 \ln 2 (0.47 + o(1))$$

implying that

$$\frac{2^{k'}}{k'}! \cdot EY_{[k']}^{\mathcal{F}} \gg e^{g_\mathcal{F}(k')} = 1.$$
As above, we get that there exists $\mu' = \mu'(k)$ and $\mathcal{F}_k^{\mu'}$ containing all isomorphic copies of $\mu'$ non-isomorphic graphs on $[k]$ such that

$$\mathbb{E}Y_{[k']^{\mu'}} = \ln 2 + o(1).$$

There exists a FO sentence describing \{\(Y_{[k']^{\mu'}} = 0\)\} with

$$\lfloor k' \rfloor + 3 \leq \lfloor k + 0.6 \rfloor + 3 \leq \lceil k \rceil + 4$$

variables. The inequalities follow from the fact that $g_{\mathcal{F}^*}(\hat{k} + 0.6) \to \infty$. We compute the second moment of $Y_{[k']^{\mu'}}$ in Section 2.2.3.

2.1.3 $\Delta \leq 0.05$ or $\Delta > 0.95$

Finally, let either $\Delta \leq 0.05$ or $\Delta > 0.95$ and $Z_k^\mathcal{F}$ be the total number of tuples $(v_1, v_2, v_3, S)$, where $v_1, v_2, v_3 \in [n]$ are distinct, $S \subseteq \binom{[n]}{3} \setminus \{v_1, v_2, v_3\}$ such that $G_n[S]$ is equal to a graph from $\mathcal{F}_k$, and there are no vertices $x$ outside $S \cup \{v_1, v_2, v_3\}$ adjacent to all vertices from $S$ and either to $v_3$ or to both $v_1, v_2$. We have

$$EZ_k^\mathcal{F} = 3 \left(\begin{array}{c} n \\ 3 \end{array} \right) \left(\begin{array}{c} n - 3 \\ k \end{array} \right) \frac{\left|\mathcal{F}_k\right|}{2^{\binom{k}{2}}} \left(1 - \frac{5}{2k + 3}\right)^{n-k-3} = e^{h_{\mathcal{F}}(k)},$$

$$h_{\mathcal{F}}(k) = (k + 3) \ln n - k \ln k + k + \ln |\mathcal{F}_k| - \frac{k(k - 1)}{2} \ln 2 - 5n2^{-k-3} + O(\ln k).$$

As above, letting $k'' \in \mathbb{R}$ be such that $f_{\mathcal{F}}(k''^*) = 0$, $\mathcal{F}_k = \mathcal{F}_k^*$, $k'' = k''^*$, $\Delta'' := k'' - \lfloor k'' \rfloor$, we get that

$$\Delta'' = k'' - \lfloor k'' \rfloor = \Delta + \log_2 \frac{5}{4} + o(1) - \left[\Delta + \log_2 \frac{5}{4} + o(1)\right] \in (0.25, 0.4)$$

implying

$$EZ_{[k'']}^\mathcal{F} = o(1), \quad \frac{2^{\binom{k''}{2}}}{k''^!}EZ_{[k'']}^\mathcal{F} \gg e^{h_{\mathcal{F}}(k'')} = 1$$

and the existence of $\mu'' = \mu''(k)$ and $\mathcal{F}_k^{\mu''}$ containing all isomorphic copies of $\mu''$ non-isomorphic graphs on $[k]$ such that

$$EZ_{[k'']}^\mathcal{F} = \ln 2 + o(1).$$

There exists a FO sentence describing \{\(Z_{[k'']}^{\mathcal{F}^{\mu''}} = 0\)\} with

$$\lfloor k'' \rfloor + 4 = \lfloor k'' - 0.25 \rfloor + 4 \leq \lceil \hat{k} - 0.09 \rceil + 4 \leq \lceil k \rceil + 4$$

variables. The inequalities follow from the fact that $h_{\mathcal{F}^*}(\hat{k} + 0.16) \to \infty$. We compute the second moment of $Z_{[k'']}^{\mathcal{F}^{\mu''}}$ in Section 2.2.4.
2.2 The second moment and \( \varphi \)-mixing

Let us recall that the \( m \)th factorial moment of a random variable \( \xi \) is \( \mathbb{E} \xi^{(m)} = \mathbb{E} \xi(\xi-1) \ldots (\xi-m+1) \) and that, for \( \xi \sim \text{Pois}(\lambda) \), \( \mathbb{E} \xi^{(m)} = \lambda^m \). It is well known that convergences of moments imply the convergence in distribution (in particular, in the case of the convergence to a Poisson random variable), see, e.g., \cite[Theorem 30.2]{2}. Therefore, in order to prove the first part of Theorem 1, it is sufficient to show that the \( m \)th factorial moment of a random variable, see, e.g., \cite[Theorem 30.2]{2}. Therefore, in order to prove the first part of Theorem 1, it is sufficient to show that the second factorial moment approaches \( (\ln 2)^2 \) while computing \( m \)th moments for \( m \geq 3 \) is not so evident. So, instead of computing all moments, we bound the probability of non-existence from below and above using \cite[Lemmas 4.1, 4.2]{9}. Let us recall these bounds.

Let \( A_1, \ldots, A_N \) be arbitrary events, and \( X = \sum_{i=1}^{N} I[A_i] \) count the number of events that occur. Consider an arbitrary directed graph \( D \) on the vertex set \([N]\), and define

\[
\psi_1 = \sum_{i=1}^{N} \sum_{(j,i) \in D, j < i} P(A_i \cap A_j), \quad \psi_2 = \sum_{i=1}^{N} \sum_{(j,i) \in D, j < i} P(A_i)P(A_j),
\]

\[
\varphi = \max_i |P(U_{(j,i) \notin D, j < i} A_i) - P(U_{(j,i) \notin D, j < i} A_j)|.
\]

In \cite{9}, it is proven that if \( \psi_1, \varphi = o(1) \), then \( P(X = 0) = \prod_{i=1}^{N} (1 - P(A_i)) + o(1) \).

Let \( X := X_{[k]}^\mu \), \( Y := Y_{[k']}^\mu \), \( Z := Z_{[k'']}^\mu \). Note that \( X = \sum_{i \in [(n)]} X_i \), where \( X_i = I[A_i^X] \) is Bernoulli random variable equals to 1 if the event \( A_i^X \) holds;

\[
A_i^X = \{ \text{ith } k\text{-set } S_i \text{ induces a copy of a graph from } \mathcal{F}^\mu \text{ in } G_n, \]

and there are no vertices outside \( S_i \) adjacent to all vertices of \( S_i \} \).

In the same way, we define \( Y = \sum_{i \in [(n-2)]} Y_i \), \( Z = \sum_{i \in [(n-3)]} Z_i \), where \( Y_i = I[A_i^Y] \), \( Z_i = I[A_i^Z] \).

\[
A_i^Y = \{ \text{ith } k\text{-tuple } (v_1^i, v_2^i, S_i) \text{ is such that } S_i \text{ induces a graph from } \mathcal{F}^\mu \text{ in } G_n, \]

and there are no \( u \notin S_i \sqcup \{ v_1^i, v_2^i \} \) adjacent to all vertices of \( S_i \) and at least 1 vertex among \( v_1^i, v_2^i \} \).

\[
A_i^Z = \{ (v_1^i, v_2^i, v_3^i, S_i) \text{ is such that } S_i \text{ induces a graph from } \mathcal{F}^\mu \text{ in } G_n, \]

and no \( u \notin S_i \sqcup \{ v_1^i, v_2^i, v_3^i \} \) is adjacent to all vertices of \( S_i \) and to either \( v_3^i \) or to both \( v_1^i, v_2^i \} \).

We define \( D^X, D^Y, D^Z \) in the most naive way: \( (i, j) \in E(D^{(i)}) \), if the \( i \)th and \( j \)th tuples are distinct and overlap. For these tuples of events we define \( (\psi_1^X, \psi_2^X, \varphi^X), (\psi_1^Y, \psi_2^Y, \varphi^Y), (\psi_1^Z, \psi_2^Z, \varphi^Z) \) as suggested above. If we show that all these parameters approach 0, then we immediately get \( P(X = 0) = e^{-EX(1+o(1))} = \frac{1}{2} + o(1) \), and the same holds for \( Y \) and \( Z \). Let us first show that \( \varphi^X, \varphi^Y, \varphi^Z = o(1) \).
2.2.1 $\varphi$-mixing

Note that $P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \bigg| A^X_i \right) - P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \right)$ is bounded from above by the probability that there exists a set $S_j$ such that

- $(j, i) \not\in D^X, j < i$,
- $S_j$ induces a copy of a graph from $\mathcal{F}^\mu$,
- there are no vertices outside $S_i \cup S_j$ adjacent to all vertices of $S_j$,
- and there is a vertex from $S_i$ that is adjacent to all vertices from $S_j$.

Therefore,

$$P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \bigg| A^X_i \right) - P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \right) \leq EX(1 - (1 - 2^{-k})^k)(1 + o(1)) \to 0. \quad (3)$$

On the other hand,

$$P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \bigg| A^X_i \right) = \frac{P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \cap A^X_i \right)}{P(A^X_i)}.$$

By construction, the events $\bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j$ and $A^X_i$ are monotone decreasing. Therefore, by FKG inequality [10, Theorem 2.12],

$$P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \right) \geq P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \right) P(A^X_i)$$

implying that

$$P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \bigg| A^X_i \right) \geq P\left( \bigcup_{(j,i) \not\in D^X \cap j<i} A^X_j \right). \quad (4)$$

Combining (3) with (4), we get that $\varphi^X = o(1)$ as needed.

The bounds for $\varphi^Y$ and $\varphi^Z$ are the same: the lower bound is immediate since $A^Y_i, A^Z_i$ are decreasing events. The upper bounds are similar to (3). Indeed, $P\left( \bigcup_{(j,i) \not\in D^Y \cap j<i} A^Y_j \bigg| A^Y_i \right) - P\left( \bigcup_{(j,i) \not\in D^Y \cap j<i} A^Y_j \right)$ is bounded from above by the probability that there exists a tuple $(v^1_j, v^2_j, S_j)$ such that

- $(j, i) \not\in D^Y, j < i$,
- $S_j$ induces a copy of a graph from $\mathcal{F}^{\nu'}$,
- there are no vertices outside $\{v^1_j, v^2_j, v^1_i, v^2_i\} \sqcup S_i \sqcup S_j$ adjacent to all vertices of $S_j$ and at least 1 vertex among $v^1_j, v^2_j$,
- there is a vertex from $S_i \sqcup \{v^1_i, v^2_i\}$ that is adjacent to all vertices from $S_j$ and at least 1 vertex among $v^1_j, v^2_j$. 

8
Then
\[
P\left( \bigcup_{(j,i) \notin \mathcal{D}_r : j < i} A_j^2 \mid A_i^2 \right) - P \left( \bigcup_{(j,i) \notin \mathcal{D}_r : j < i} A_j^2 \right) \leq EY \left( 1 - (1 - 3 \cdot 2^{-k-2})^{k+2} \right) (1 + o(1)) \to 0.
\]
The proof of the upper bound
\[
P\left( \bigcup_{(j,i) \notin \mathcal{D}_r^z : j < i} A_j^2 \mid A_i^2 \right) - P \left( \bigcup_{(j,i) \notin \mathcal{D}_r^z : j < i} A_j^2 \right) \leq EZ \left( 1 - (1 - 5 \cdot 2^{-k-3})^{k+3} \right) (1 + o(1)) \to 0
\]
is identical.

### 2.2.2 The second moment of $X$

We here compute the second moment of $X := X_{[k]}^{d^n}$ or, more precisely, show that $\psi_1^X = o(1)$ and $\psi_2^X = o(1)$. We have

\[
\psi_1^X \leq \sum_{s=1}^{k-1} \binom{n}{k} \binom{k}{s} \binom{n-s}{k-s} \frac{|F_k^\mu|^2}{2^{k/2} \binom{k}{s}} \left( 1 - \frac{1}{2^s} + \frac{1}{2} \left( 1 - \frac{1}{2^{k-s}} \right)^2 \right) ^{n-2k+s}
\]

\[
\leq \left( \frac{n}{k} \right)^2 \frac{|F_k^\mu|^2}{2^{k/2} \binom{k}{s}} \sum_{s=1}^{k-1} \frac{k^s}{s!} \frac{k^s}{n^s} \left( 1 - \frac{2}{2^k} + \frac{1}{2^{2k-s}} \right) ^{n-2k+s}
\]

\[
\leq (EX)^2 (1 + o(1)) \sum_{s=1}^{k-1} \left( \frac{ek^2}{sn} \frac{s-1}{s} \right)^s \exp \left[ \frac{n}{2^{2k-s}} \right].
\]

There is $s_0 = k(1 - o(1))$ such that $\frac{n}{2^{2k-s}} = o(1)$ for all $s \leq s_0$. If $s_0 < s \leq k - 1$, then $\frac{n}{2^{2k-s}} = \frac{k^2 \ln n}{4} (1 + o(1))$. Therefore,

\[
\frac{\psi_1^X}{(EX)^2} \leq (1 + o(1)) \left( \sum_{s=1}^{s_0} \left( \frac{ek^2}{sn} \frac{s-1}{s} \right)^s \right.
\]

\[
+ \sum_{s=s_0+1}^{k-1} \exp \left[ -s \ln n + \frac{s^2}{2} \ln 2 \right] (1 + o(1)) \left( \frac{k^2 \ln 2}{4} (1 + o(1)) \right) \]

\[
\leq \sum_{s=1}^{s_0} \left( \frac{k^2}{2^{k/2}} \right)^s + \sum_{s=s_0+1}^{k-1} \exp \left[ - (1 + o(1)) \frac{k^2 \ln 2}{2} + \frac{k^2 \ln 2}{4} (1 + o(1)) \right] = o(1).
\]

Moreover,

\[
\psi_2^X = \sum_{s=1}^{k-1} \binom{n}{k} \binom{k}{s} \binom{n-k}{k-s} \left( \frac{|F_k^\mu|^2}{2^{k/2} \binom{k}{s}} \left( 1 - \frac{1}{2^k} \right) ^{n-k} \right)^2
\]

\[
= (EX)^2 \sum_{s=1}^{k-1} \binom{k}{s} \binom{n-k}{k-s} \left( \frac{ek^2}{sn} \right)^s \leq (EX)^2 \sum_{s=1}^{k-1} \left( \frac{ek^2}{sn} \right)^s = o(1).
\]
Note that, since $A_i^X$ are monotone decreasing, by FKG inequality, we have that $P(A_i^X \cap A_j^{X'}) \geq P(A_i^X)P(A_j^{X'})$ for all $i,j$. Therefore, $\phi_2^X \leq \phi_1^X$, and the same holds for $Y$ and $Z$. Therefore, we are actually not obliged to compute $\phi_2$ — it is sufficient to prove that $\phi_1 = o(1)$.

In the next two sections, we show that $\phi_1^Y = o(1)\ , \phi_2^Z = o(1)$, and this finishes the proof of the first part of Theorem 11.

### 2.2.3 The second moment of $Y$

Let $F$ be the family of graphs on $[k+2]$ that is obtained from $F_k^\mu$ by transforming each $F \in F_k^\mu$ into all possible $2^{k+2}$ graphs on $[k+2]$ containing $F$ as a subgraph.

We separately count contributions to $\psi_1^Y$ of pairs of overlapping tuples that have at most $k$ vertices in common, and pairs of tuples that have $k+1$ vertices in common. For any two tuples $(v_i^1, v_i^2, S_i)$ and $(v_j^1, v_j^2, S_j)$ that share $s \leq k$ vertices and a vertex $z \notin \{v_i^1, v_i^2, v_j^1, v_j^2\} \cup S_i \cup S_j$, the probability that $z$ is adjacent to all vertices of $S_i \cup S_j$ at least 1 vertex among $v_i^1, v_i^2$ and at least 1 vertex among $v_j^1, v_j^2$, is at most $\frac{9}{2^{2k+4-s}}$. Moreover, if $s = k+1$, then this probability can be bounded by $\frac{5}{2^{k+3}}$ since its maximum is achieved when $S_i = S_j$, $v_i^2 = v_j^2$, and $v_i^1 \neq v_j^1$. In this case, there are exactly 5 ways to draw edges from $z$ to $(v_i^1, v_i^2, S_i)$ and $(v_j^1, v_j^2, S_j) \cup S_j$ that satisfies the mentioned restrictions.

Therefore, we get

$$\psi_1^Y \leq \sum_{s=1}^{k} \binom{n}{k+2} \binom{k+2}{s} \binom{n-k-2}{k+2-s} \binom{k+2}{2} \frac{|F|^2}{2^{2(k+2)-(k-s)}} \left(1 - 2 \cdot \frac{3}{2^{k+2}} + \frac{9}{2^{2k+4-s}}\right)^{s-2(k+1)}$$

$$+ \left(\frac{n}{k+2}\right)(k+2)(n-k-2) \binom{k+2}{2} \frac{|F|^2}{2^{2(k+2)-(k+1)}} \left(1 - 2 \cdot \frac{3}{2^{k+2}} + \frac{5}{2^{k+3}}\right)^{n-k-3}$$

$$\leq ((EY)^2 + o(1)) \left(\frac{\sum_{s=1}^{k} \binom{k}{2} \binom{n-k-2}{k+2-s} \binom{k+2}{2} \frac{9n}{2^{2k+4-s}}}{n^{k+2}}\right)$$

$$+ \frac{(k+2)(n-k-2)}{n^{k+2}} \frac{2^{(k+1)}}{2^{2(k+1)}} \left(\frac{9n}{2^{2k+4-s}}\right)$$

$$\leq ((EY)^2 + o(1)) \left(\frac{\sum_{s=1}^{k} \binom{k}{2} \binom{n-k-2}{k+2-s} \binom{k+2}{2} \frac{9n}{2^{2k+4-s}}}{n^{k+2}}\right) + \frac{(k+2)!}{n^{k+2}} \frac{2^{(k+1)}}{2^{2(k+1)}} \left(\frac{9n}{2^{2k+4-s}}\right)$$

There is $s_0 = k(1-o(1))$ such that

$$\frac{n}{2^{2k+4-s}} = o(1) \text{ for all } s \leq s_0.$$

If $s_0 < s \leq k$, then

$$\frac{9n}{2^{2k+4-s}} \leq \frac{3}{4} \cdot \frac{3n}{2^{k+2}} = \frac{3k^2 \ln 2}{8}(1 + o(1)).$$

In the same way,

$$\frac{5n}{2^{k+3}} = \frac{5k^2 \ln 2}{12}(1 + o(1)).$$
Therefore,

\[
\frac{\psi^Y}{(EY)^2} \leq (1 + o(1)) \sum_{s=1}^{s_0} \left( \frac{e(k+2)^2}{sn} \right)^s \left( \frac{2^{s-1}}{2^{s-1}} \right) + \exp \left[ \left( -k \ln n + \frac{k^2}{2} \ln 2 \right) \left( 1 + o(1) \right) + \frac{5k^2 \ln 2}{12} \left( 1 + o(1) \right) \right]
+ \exp \left[ \left( -s \ln n + \frac{s^2}{2} \ln 2 \right) \left( 1 + o(1) \right) + \frac{3k^2 \ln 2}{8} \left( 1 + o(1) \right) \right]
\leq (1 + o(1)) \sum_{s=1}^{s_0} \left( \frac{k^2}{2k/2} \right)^s + \sum_{s=s_0+1}^k \exp \left[ - \left( 1 + o(1) \right) \frac{k^2 \ln 2}{8} \right]
+ \exp \left[ - \left( 1 + o(1) \right) \frac{k^2 \ln 2}{12} \right] = o(1).
\]

### 2.2.4 The second moment of \( Z \)

Let \( \mathcal{F} \) be the family of graphs on \([k+3]\) that is obtained from \( \mathcal{F}^{\mu''}_k \) by transforming each \( F \in \mathcal{F}^{\mu''}_k \) into all possible \( 2^{3+3k} \) graphs on \([k+3]\) containing \( F \) as a subgraph.

For any two tuples \((v^1_i, v^2_i, v^3_i, S_i)\) and \((v^1_j, v^2_j, v^3_j, S_j)\) that share \( s \leq k \) vertices and a vertex \( z \notin \{v^1_i, v^2_i, v^3_i, v^1_j, v^2_j, v^3_j\} \cup S_i \cup S_j \), the probability that \( z \) is adjacent to all vertices of \( S_i \cup S_j \), either to \( v^3_i \), or to both \( v^1_i, v^2_i \), and either to \( v^3_j \), or to both \( v^1_j, v^2_j \), is at most \( 25 \). Moreover, if \( s = k + 1 \), then this probability can be bounded by \( \frac{25}{2^{k+5}} \) since its maximum is achieved when \( S_i = S_j, v^3_i = v^3_j, v^1_i, v^2_i, v^1_j, v^2_j \) are all distinct. Finally, if \( s = k + 2 \), then this probability can be bounded by \( \frac{9}{2^{k+5}} \) since its maximum is achieved when \( S_i = S_j, v^3_i = v^3_j, v^1_i = v^1_j \), and \( v^1_i \neq v^1_j \).

We have

\[
\psi^Z \leq \sum_{s=1}^k 9 \left( \binom{n}{k+3} \binom{k+3}{s} \binom{n-k-3}{k+3-s} \binom{k+3}{3} \right)^2 \frac{|\mathcal{F}|^2}{2^{(k+3)-(s/2)}} \left( 1 - 2 \cdot \frac{5}{2^{k+3}} + \frac{25}{2^{2k+6-s}} \right)^{n-2k-6+s}
+ \sum_{s=s_0+1}^k \exp \left[ - \left( 1 + o(1) \right) \frac{k^2 \ln 2}{8} \right]
\leq (EZ)^2 + o(1) \left( \sum_{s=1}^k \frac{\binom{k+3}{s} \binom{n-k-3}{k+3-s} \binom{n}{k+3}}{\binom{n}{k+3}} \right) \binom{2}{s} \exp \left[ \frac{25n}{2^{k+5}} \right]
+ \frac{\binom{k+3}{2} \binom{n-k-3}{k+3} \binom{n}{k+3}}{\binom{n}{k+3}} \binom{2}{s} \exp \left[ \frac{17n}{2^{k+5}} \right] + \frac{\binom{k+3}{2} \binom{n-k-3}{k+3} \binom{n}{k+3}}{\binom{n}{k+3}} \binom{2}{s} \exp \left[ \frac{9n}{2^{k+4}} \right]
\]
\[
\leq ((EZ)^2 + o(1)) \left( \sum_{s=1}^{k} \left( \frac{e(k+3)^2}{sn} \right)^s \cdot 2^{s-1} \right) \exp \left[ \frac{25n}{2^{k+6-s}} \right] + \frac{(k+3)!}{n^k} 2^{\left( \frac{k+1}{2} \right)} \exp \left[ \frac{17n}{2^{k+5}} \right] + \frac{(k+3)!}{n^{k+1}} 2^{\left( \frac{k+2}{2} \right)} \exp \left[ \frac{9n}{2^{k+4}} \right].
\]

As above we find \( s_0 = k(1 - o(1)) \) such that \( \frac{n}{2^{k+6-s}} = o(1) \) for all \( s \leq s_0 \). Then

\[
\frac{\psi Z}{(EZ)^2} \leq (1 + o(1)) \sum_{s=1}^{s_0} \left( \frac{e(k+3)^2}{sn} \right)^s \cdot 2^{s-1} \exp \left[ \left( -k \ln n + \frac{k^2}{2} \ln 2 + \frac{9k^2 \ln 2}{20} \right) \right] + \exp \left[ \left( -k \ln n + \frac{k^2}{2} \ln 2 + \frac{17k^2 \ln 2}{40} \right) \right] + \sum_{s=s_0+1}^{k} \exp \left[ \left( -k \ln n + \frac{k^2}{2} \ln 2 + \frac{5k^2 \ln 2}{16} \right) \right] \leq (1 + o(1)) \sum_{s=1}^{s_0} \left( \frac{k^2}{2^{k/2}} \right)^s \exp \left[ -\left( 1 + o(1) \right) \frac{k^2 \ln 2}{20} \right] = o(1).
\]

3 Proof of Theorem 1 part 2

The proof strategy is very similar. The difference is that, instead of finding a monotone isomorphism-closed family of graphs such that the respective expected number is asymptotically a constant, we find a set \( S_1 \) in one of the random graphs (say, \( G_1 \)) that is ‘non-extendible’ in some sense, and show that in the other random graph whp any set \( S_2 \) that induces a subgraph isomorphic to \( G_1[S] \) is extendible.

Consider a uniformly distributed random graph \( G_n \) on \([n]\), and let \( X_k \) be the total number of \( k \)-sets \( S \subset [n] \) such that there is no vertex \( x \) outside \( S \) adjacent to all vertices in \( S \). Then

\[
EX_k = \binom{n}{k} \left( 1 - \frac{1}{2^k} \right)^{n-k} = e^{f(k)}, \quad \text{where} \quad f(k) = k \ln n - k \ln k + k - n2^{-k} + O(\ln k).
\]

Let \( k \in \mathbb{R} \) be such that \( f(k) = 0 \). Note that \( EX_{[k]} \leq 1 \) and \( EX_{[k]} \geq 1 \). Let \( \Delta := [k] - k \). Then

\[
f([k]) - f(k) = k^2 \ln 2 \left( 1 - 2^{-\Delta} + o(1) \right).
\]

Note that we need \( EX_{[k]} \to \infty \) in order to find in the first random graph a non-extendible \([k]-set \). Also, we need \([k]!2^{(\frac{[k]}{2})} [k]!2^{(\frac{[k]}{2})} \to 0 \) in order to show that in the second random graph there is no non-extendible \([k]-set \) that induces a graph isomorphic to the graph induced by \( S \) in the first random graph.

In order to be certain that \( EX_{[k]} \to \infty \), we need \( 1 - 2^{-\Delta} \) to be bounded away from 0. Moreover, in order to have \([k]!2^{(\frac{[k]}{2})} \to 0 \), we need \( 1 - 2^{-\Delta} \) not to be very large. Let us
first consider the sufficient restriction $0.05 \leq \Delta \leq 0.95$. For other values of $\Delta$, we will slightly modify the sentence in the same way as it was done in Section 2.1.2.

Since $1 - 2^{-\Delta}$ is bounded away both from 0 and from 1/2, we get that $EX_{[k]} \to \infty$ and $[k]!2^{(-[k]})EX_{[k]} \to 0$.

On the one hand, it means that $P(X_{[k]} > 0) \to 1$. Indeed, to show this, it is sufficient (by Chebyshev’s inequality) to prove that $EX_{[k]}(X_{[k]} - 1) = (1 + o(1))(EX_{[k]})^2$. Note that this second factorial moment equals to the summation over all choices of pairs of distinct $[k]$-set $S_1, S_2$ of the probability $P_{t,j}$ that each of the sets does not have a vertex outside adjacent to all vertices inside. In Section 2.2.2, we showed that the contribution of intersecting pairs is $o((EX_{[k]})^2)$. The contribution of non-intersection pairs is at most

$$\left(\frac{n}{[k]}\right)^2 \left(1 - \frac{2}{[k]}\right)^{n-2[k]} \sim (EX_{[k]})^2$$
as needed.

On the other hand, for a given graph $F$ on $[k]$, by the union bound, the probability that there is a set $S$ in $G(n, p)$ that induces a subgraph isomorphic to $F$ such that every vertex outside $S$ is not adjacent to at least 1 vertex of $S$ is at most

$$\left(\frac{n}{[k]}\right)[k]! \left(\frac{[k]}{2^{([k]}}}\right) \left(1 - \frac{1}{2[k]}\right)^{n-[k]} \to 0.$$

Note that $k = \hat{k} + o(1)$ (recall that $\hat{k}$ is defined in Section 11 as the unique root of the equation $f_{\text{ext}}(k) = 0$) implying (since $\Delta \geq 0.05$) that $[k] = [\hat{k}] = [\hat{k}] + 1$. The sentence requires $[k] + 1 \leq [\hat{k}] + 2$ variables.

Finally, let either $\Delta < 0.05$, or $\Delta > 0.95$. Let $Y$ be the total number of tuples $(v_1, v_2, S)$, where $v_1 \neq v_2 \in [n]$, $S \in ([n] \setminus \{v_1, v_2\})$, such that there are no vertices $x$ outside $S \cup \{v_1, v_2\}$ adjacent to all vertices from $S$ and at least 1 vertex among $v_1, v_2$. We have

$$EY_{\hat{k}} = \binom{n}{2} \binom{n-2}{k} \left(1 - \frac{3}{2^{k+2}}\right)^{n-k-2} = e^{g(k)},$$

$$g(k) = (k + 2) \ln n - k \ln k + k - 3n2^{-k-2} + O(\ln k).$$

Let $k' \in \mathbb{R}$ be such that $g(k') = 0$. Note that $k' = k + \log_2 3 - 2 + o(1)$ and

$$\Delta' := k' - [k'] = \Delta + \log_2 \frac{3}{2} + o(1) - \left[\Delta + \log_2 \frac{3}{2} + o(1)\right] \in (0.5, 0.65).$$

Then $EY_{[k']} \to \infty$ and $[k']!2^{(-[k'])}EY_{[k']} \to 0$ as needed. In the same way as above, applying the bounds from Section 2.2.3 and Chebyshev’s inequality, we get that $P(Y_{[k']} > 0) \to 1$. Also, for a given graph $F$ on $[[k']]$, by the union bound, the probability that there are a set $S$ and a
pair of distinct vertices $v_1, v_2$ outside $S$ in $G(n, p)$ such that $S$ induces a subgraph isomorphic to $F$ and every common neighbor $u \notin \{v_1, v_2\}$ of $S$ is not adjacent to any of $v_1, v_2$ is at most

$$\binom{n}{2} \left( \frac{n-2}{k} \right) \frac{[k]!}{2^{|k|}} \left( 1 - \frac{3}{2^{|k|+2}} \right)^{n-[k]-2} \to 0.$$ 

This finishes the proof of part 2 of Theorem 1 since the number of variables we need is $[k'] + 3 = [k + \log_2 3 - 2 + o(1)] + 3 = [\hat{k}] + 2 \leq \hat{k} + 3$.

4 Other logics and models of random graphs

Let us note that our results are immediately extended to infinite FO logic with finitely many variables [12, Section 11.1] since two graphs that satisfy the extension axiom $\varphi_k$ are not distinguishable in the fragment of this logic with $k + 1$ variables as well [12, Proposition 11.8]. On the other hand, there are monadic second order sentences such that their truth probability does not converge on $G_n$ to 0 or 1. Actually, 1 existential monadic quantifier and 2 FO variables is enough [13] to distinguish between two independent uniformly distributed random graph on $[n]$ at least for some sequences of $n$ (note that in the cited paper it is proven that there exist a sentence $\varphi$ and two sequences $n_i, m_i$ such that $P(G_{n_i} \models \varphi) \to 0$ and $P(G_{m_i} \models \varphi)$ is bounded away from 0; however, it is easy to see that the same method can be applied to show that there exists a sequence $n_i$ such that $P(G_{n_i} \models \varphi)$ is bounded away both from 0 and 1 for the same sentence $\varphi$). It would be interesting to know if the same number of variables is sufficient for all large enough $n$ to distinguish between two independent random graphs on $[n]$. It would also be of interest to study distinguishability of random graphs with other probability distributions. In particular, the classical 0-1 law [5, 7] can be generalized to the binomial random graph $G(n, p)$ for a wide range of edge probabilities $p$ (note that $p = 1/2$ corresponds to the considered in the paper uniform distribution): if $np^\alpha \to \infty$ for all $\alpha > 0$, then $G(n, p)$ obeys FO 0-1 law [17]. Our results can be also generalized to the same range of $p = p(n)$. In particular, for every constant $p \in (0, 1)$, we will also get concentration of the minimum number of variables required to distinguish between two random graphs in a set of constantly many numbers. Note that, in [11], it is proven that, for $p = 1/2$, there exists a sequence $\{n_i\}_{i \in \mathbb{N}}$ such that whp the minimum quantifier depth (and, therefore, the number of variables) of a FO sentence defining $G(n_i, 1/2)$ is concentrated in a set of 5 consecutive numbers. However, it is not clear if there is a constant-size concentration interval for the whole sequence of $n$ and also for other constant $p \in (0, 1)$.

The case $p = n^{-\alpha}$ is different. When $\alpha$ is rational and belongs to $(0, 1]$ or $\alpha = 1 + \frac{1}{m}$ for some $m \in \mathbb{N}$, then $G(n, p)$ does not obey FO 0-1 law [16] since, for any such $\alpha$, there exists a constant size strictly balanced graph with density $1/\alpha$ [14], and the number of isomorphic copies of such a graph in $G(n, n^{-\alpha})$ converges in distribution to a Poisson random variable (see [10, Theorem 3.19]). Therefore, with probability bounded away from 0, two independent copies of $G(n, p)$ can be distinguished by a FO sentence with constant number of variables.
(equal to the number of vertices in the mentioned strictly balanced graph). It is also not hard to see that, for every $\varepsilon > 0$, there exists a constant $k = k(\varepsilon)$ such that two independent random graphs are $k$-distinguishable with probability at least $1 - \varepsilon$, since the joint distribution of the numbers of copies of different strictly balanced graphs with the same density $1/\alpha$ converges to a vector of independent Poisson random variables (see [10, Remark 3.20]). It is not so clear how to construct a fix sentence with constantly many variables that distinguishes between two independent random graphs with probability arbitrarily close to $1 - \varepsilon$. However, we believe that this is possible and boils down to constructing strictly balanced graphs with fixed densities and linearly (in the number of vertices) growing number of automorphisms.

If $\alpha > 1$ and $\alpha \neq 1 + \frac{1}{m}$ for any $m \in \mathbb{N}$, then whp $G(n, p = n^{-\alpha})$ is a forest such that every tree with at most $\left\lfloor \frac{1}{\alpha} - 1 \right\rfloor$ vertices appears as a component in $G(n, p)$ at least $n^\varepsilon$ times (for sufficiently small constant $\varepsilon > 0$), and there are not trees of size $\left\lfloor \frac{1}{\alpha} - 1 \right\rfloor$ (see, e.g., [10, Theorems 3.4, 3.29]). Moreover, the number of components isomorphic to a fixed small tree satisfies the central limit theorem [15]. From this, it is not hard to derive that $n^\varepsilon$ variables is needed to distinguish between two independent copies of $G(n, p)$, and that this number is not concentrated in any bounded interval whp.

The most intriguing case is when $\alpha \in (0, 1)$ is irrational. Then the random graph $G(n, p = n^{-\alpha})$ obeys FO 0-1 law [16]. So, any fixed sentence does not distinguish between two independent random graphs whp. In the same way as in the case of rational $\alpha$, one may consider an infinite sequence of strictly balanced graphs $H_i$, $i \in \mathbb{N}$, such that the density $\rho(H_i) = \frac{|E(H_i)|}{|V(H_i)|}$ approaches $\frac{1}{\alpha}$ as $i \to \infty$. However, in order to make the expected number of $H_i$ to be asymptotically a constant, we need $|V(H_i)| = \Omega(\ln n)$. Nevertheless, we conjecture that for every $\varepsilon > 0$ and every $\frac{1}{2} < p < n^{-\varepsilon}$, the minimum number of variables of a FO sentence that distinguishes between two independent copies of $G(n, p)$ is $o(\ln n)$.

Finally, let us mention other random graph models, in particular random $d$-regular graphs $G_{n, d}$. Note that, in the dense case (say, $d = \frac{1}{2}n(1 + o(1))$), FO 0-1 law holds [8]. However, our results can not be immediately translated using, say, sandwiching techniques [6] since we need the coupling $G(n, p_1) \subset G_{n, d} \subset G(n, p_2)$ with $p_1, p_2$ very close to $d$ in order to translate the results, but this is not likely to be true. However, we believe that an analogue of Theorem 1 holds true for $G_{n, d}$ as well. Let us also mention the sparse case of constant degree $d$: two random $d$-regular graphs can be distinguished with constantly many variables since there is no FO 0-1 law due to the convergence to Poisson distribution of the number of constant size cycles [18]. On the other hand, for sufficiently small $\varepsilon > 0$, it is possible to construct 3-regular non-isomorphic graphs on $n$ vertices that can not be distinguished by at most $\varepsilon n$ variables in the first order logic with counting quantifiers (that is even stronger) [3]. It is natural to ask the same question about Cayley graphs. Do there exist 3-regular non-isomorphic Cayley graphs on $n$ vertices that can not be distinguished by at most $\varepsilon n$ variables in the first order logic with counting quantifiers? Note that whp 4 variables is enough to distinguish between two independent uniformly distributed random graphs by a FO sentence with counting [1].
References

[1] L. Babai, L. Kučera, *Canonical labeling of graphs in linear average time*, Proc. 20th IEEE Symp. Found. Computer Sci., 1979, pp. 39–46.

[2] P. Billingsley, *Probability and Measure*, Third Edition, J. Wiley & Sons, 1995.

[3] J.-Y. Cai, M. Fürer, N. Immerman, *An optimal lower bound on the number of variables for graph identification*, Combinatorica 12 (1992) 389–410.

[4] P. Erdős and A. Rényi, *Asymmetric graphs*, Acta Mathematica Hungarica, 14:3-4 (1963) 295–315.

[5] R. Fagin, *Probabilities in finite models*, J. Symbolic Logic 41 (1976) 50–58.

[6] P. Gao, M. Isaev, B. D. McKay, *Sandwiching random regular graphs between binomial random graphs*, SIAM Conference on Discrete Algorithms (SODA 2020) 690–701.

[7] Y.V. Glebskii, D.I. Kogan, M.I. Liogon’kii, V.A. Talanov, *Range and degree of realizability of formulas the restricted predicate calculus*, Cybernetics 5 (1969) 142–154 (Russian original: Kibernetica 2, 17–27).

[8] S. Haber, M. Krivelevich, *The logic of random regular graphs*, Journal of combinatorics 1:4 (2010) 389–440.

[9] M. Isaev, I. Rodionov, R. Zhang, M. Zhukovskii, *Extremal independence in discrete random systems*, arXiv:2105.04917.

[10] S. Janson, T. Luczak, A. Rucinski, *Random graphs*, J. Wiley & Sons, 2000.

[11] J.H. Kim, O. Pikhurko, J.H. Spencer, O. Verbitsky, *How complex are random graphs in first order logic?*, Random structures & algorithms 26:1-2 (2005) 119–145.

[12] L. Libkin, *Elements of finite model theory*, Springer, 2004.

[13] S. Popova, M. Zhukovskii, *Existential monadic second order logic of undirected graphs: The Le Bars conjecture is false*, Annals of Pure and Applied Logic 170:4 (2019) 505–514.

[14] A. Ruciński, A. Vince, *Balanced graphs and the problem of subgraphs of a random graph*, Congressus Numerantium, 49 (1985) 181–190.

[15] A. Ruciński, *When are small subgraphs of a random graph normally distributed?*, Probability theory and related fields, 78:1 (1988) 1–10.

[16] S. Shelah, J.H. Spencer, *Zero-one laws for sparse random graphs*, J. Amer. Math. Soc., 1 (1988) 97–115.
[17] J. H. Spencer, *Threshold spectra via the Ehrenfeucht game*, Discrete Applied Math. 30 (1991) 235–252.

[18] N. Wormald, *The asymptotic distribution of short cycles in random regular graphs*, J. Combin. Theory Ser. B 31 (1981) 168–182.