

A Combinatorial Characterization of the Testable Graph Properties: It's All About Regularity *

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ABSTRACT

A common thread in recent results concerning the testing of dense graphs is the use of Szemerédi's regularity lemma. In this paper we show that in some sense this is not a coincidence. Our first result is that the property defined by having any given Szemerédi-partition is testable with a constant number of queries. Our second and main result is a purely combinatorial characterization of the graph properties that are testable with a constant number of queries. This characterization (roughly) says that a graph property \mathcal{P} can be tested with a constant number of queries **if and only if** testing \mathcal{P} can be reduced to testing the property of satisfying one of finitely many Szemerédi-partitions. This means that in some sense, testing for Szemerédi-partitions is as hard as testing any testable graph property. We thus resolve one of the main open problems in the area of property-testing, which was raised in the 1996 paper of Goldreich, Goldwasser and Ron [24] that initiated the study of graph property-testing. This characterization also gives an intuitive explanation as to what makes a graph property testable.

Categories and Subject Descriptors

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1. BACKGROUND

1.1 Basic definitions

The meta-problem in the area of property testing is the following: Given a combinatorial structure S , distinguish between the case that S satisfies some property \mathcal{P} and the case that S is ϵ -far from satisfying \mathcal{P} . Roughly speaking, a combinatorial structure is said to be ϵ -far from satisfying some property \mathcal{P} if an ϵ -fraction of its representation should be modified in order to make S satisfy \mathcal{P} . The main goal is to design randomized algorithms, which look at a very small portion of the input, and using this information distinguish with high probability between the above two cases. Such algorithms are called *property testers* or simply *testers* for the property \mathcal{P} . Preferably, a tester should look at a portion of the input whose size is a function of ϵ only. Blum, Luby and Rubinfeld [10] were the first to formulate a question of this type, and the general notion of property testing was first formulated by Rubinfeld and Sudan [34], who studied various algebraic properties such as linearity of functions.

The main focus of this paper is the testing of properties of graphs. More specifically, we focus on property testing in the dense graph model as defined in [24]. In this case a graph G is said to be ϵ -far from satisfying a property \mathcal{P} , if one needs to add and/or delete at least ϵn^2 edges to G in order to turn it into a graph satisfying \mathcal{P} . A tester for \mathcal{P} should distinguish with high probability, say $2/3$, between the case that G satisfies \mathcal{P} and the case that G is ϵ -far from satisfying \mathcal{P} . Here we assume that the tester can query some oracle whether a pair of vertices, i and j , are adjacent in the input graph G . In what follows we will say that a tester for a graph property \mathcal{P} has *one-sided error* if it accepts every graph satisfying \mathcal{P} with probability 1 (and still rejects those that are ϵ -far from \mathcal{P} with probability at

least $2/3$). If the tester may reject graphs satisfying \mathcal{P} with non-zero probability then it is said to have *two-sided error*. The following notion of efficient testing will be the main focus of this paper:

DEFINITION 1.1. (TESTABLE) *A graph property \mathcal{P} is said to be testable if there is a randomized algorithm T , that can distinguish with probability $2/3$ between graphs satisfying \mathcal{P} and graphs that are ϵ -far from satisfying \mathcal{P} , while making a number of edge queries which is bounded by some function $q(\epsilon)$ that is independent of the size of the input.*

The study of the notion of testability for combinatorial structures, and mainly the dense graph model, was introduced in the seminal paper of Goldreich, Goldwasser and Ron [24]. Graph property testing has also been studied in the *bounded-degree* model [26], and the newer *general density* model [30]. We note that in these models a property is usually said to be testable if the number of queries is $o(n)$. Following [24, 10, 34] property testing was studied in various other contexts such as boolean functions [4, 16, 19, 31], geometric objects [2, 11] and algebraic structures [10, 20, 8]. See the surveys [14, 33] for additional results and references.

1.2 The characterization project

With this abundance of results on property testing, a natural question is what makes a combinatorial property testable. In particular, characterizing the testable graph properties was considered one of the main open problems in the area of property testing, and was raised already in the 1996 paper of Goldreich, Goldwasser and Ron [24], see also [22], [9] and [25]. In this paper we obtain for the first time a characterization of the testable graph properties. We next discuss some results related to this problem.

A natural strategy toward obtaining a characterization of the testable graphs was to either prove the testability/non-testability of general families of graph properties or to obtain characterizations for special cases of testers. The main result of [24] was that a general family of so called “partition-problems” are all testable. These include the properties of being k -colorable, having a large cut and having a large clique. [25] gave a characterization of the partition-problems that can be tested with 1-sided error. They also proved that not all graph properties that are closed under edge-removal are testable. [12] studied property testing via the framework of *abstract combinatorial programs* and gave certain characterizations within this framework. [3] tried to obtain a *logical* characterization of the testable graph properties. More specifically, it was shown that every first order graph-property of type $\exists\forall$ (see [3]) is testable, while there are first-order graph properties of type $\forall\exists$ that are not testable. The main technical result of [3] was that certain abstract colorability properties are all testable. These results were generalized in [13]. In [6] it was shown that every graph property that is closed under removal of edges and vertices is testable. This result was extended in [7], where it was shown that in fact, being closed under vertex removal is already sufficient for being testable (see also [28]). [7] also contains a characterization of the graph properties that can be tested with *one-sided* error by certain restricted testers. Finally, [25] following [3], proved that a tester may be assumed to be non-adaptive (see Lemma 4.2), and [17] proved that if a graph property is testable then it is also estimable (see Theorem 4).

2. THE MAIN RESULT

Our main result in this paper gives a purely combinatorial characterization of the testable graph properties. As we have previously mentioned, the first properties that were shown to be testable in [24] were certain graph partition properties. As it turns out, our characterization relies on certain “enhanced” partition properties, whose existence is guaranteed by the celebrated regularity lemma of Szemerédi [35]. We start by introducing some standard definitions related to the regularity lemma. For a comprehensive survey about the regularity lemma the reader is referred to [27].

For every two nonempty disjoint vertex sets A and B of a graph G , we define $e(A, B)$ to be the number of edges of G between A and B . The *edge density* of the pair is defined by $d(A, B) = e(A, B)/(|A||B|)$.

DEFINITION 2.1. (γ -REGULAR PAIR) *A pair (A, B) is γ -regular, if for any two subsets $A' \subseteq A$ and $B' \subseteq B$, satisfying $|A'| \geq \gamma|A|$ and $|B'| \geq \gamma|B|$, the inequality $|d(A', B') - d(A, B)| \leq \gamma$ holds.*

Throughout the paper it will be useful to observe that in the above definition it is enough to require that $|d(A', B') - d(A, B)| \leq \gamma$ for sets $A' \subseteq A$ and $B' \subseteq B$ of sizes $|A'| = \gamma|A|$ and $|B'| = \gamma|B|$. A partition $\mathcal{A} = \{V_i \mid 1 \leq i \leq k\}$ of the vertex set of a graph is called an *equipartition* if $|V_i|$ and $|V_j|$ differ by no more than 1 for all $1 \leq i < j \leq k$ (so in particular every V_i has one of two possible sizes). The *order* of an equipartition denotes the number of partition classes (k above).

DEFINITION 2.2. (γ -REGULAR EQUIPARTITION) *We say that an equipartition $\mathcal{B} = \{V_i \mid 1 \leq i \leq k\}$ of the vertex set of a graph is γ -regular if all but at most $\gamma \binom{k}{2}$ of the pairs (V_i, V_j) are γ -regular.*

In what follows an equipartition is said to *refine* another if every set of the former is contained in one of the sets of the latter. Szemerédi’s regularity lemma can be formulated as follows.

LEMMA 2.3 ([35]). *For every m and $\gamma > 0$ there exists $T = T_{2.3}(m, \gamma)$ with the following property: If G is a graph with $n \geq T$ vertices, and \mathcal{A} is any equipartition of the vertex set of G of order at most m , then there exists a refinement \mathcal{B} of \mathcal{A} of order k , where $m \leq k \leq T$ and \mathcal{B} is γ -regular. In particular, for every m and $\gamma > 0$ there exists $T = T_{2.3}(m, \gamma)$, such that any graph with $n \geq T$ vertices has a γ -regular equipartition of order k , where $m \leq k \leq T$.*

The regularity lemma guarantees that every graph has a γ -regular equipartition of (relatively) small order. As it turns out in many applications of the regularity lemma, one is usually interested in the densities of the bipartite graphs connecting the sets V_i of the regular partitions. In fact, one important consequence of the regularity lemma is that in many cases knowing the densities connecting the sets V_i (approximately) tells us all we need to know about a graph. Roughly speaking, if a graph G has a regular partition of order k and we define a weighted graph $R(G)$, of size k , where the weight of edge (i, j) is $d(V_i, V_j)$, then by considering an appropriate property of $R(G)$ one can infer many properties of G . As the order of the equipartition is guaranteed to be bounded by a function of γ , this means that for

many applications, every graph has an approximate description of *constant-complexity* (we will return to this aspect in a moment). As it turns out, this interpretation of the regularity lemma is the key to our characterization. We believe that our characterization of the testable graph properties is an interesting application of this aspect of the regularity lemma.

Given the above discussion it seems natural to define a graph property, which states that a graph has a given γ -regular partition, that is, an equipartition which is γ -regular and such that the densities between the sets V_i belong to some predefined set of densities.

DEFINITION 2.4. (REGULARITY INSTANCE) *A regularity-instance is given by an error-parameter $0 < \gamma \leq 1$, an integer k , a set of $\binom{k}{2}$ densities $0 \leq \eta_{i,j} \leq 1$ indexed by $1 \leq i < j \leq k$, and a set \bar{R} of pairs (i, j) of size at most $\gamma \binom{k}{2}$. A graph is said to satisfy the regularity-instance if it has an equipartition $\{V_i \mid 1 \leq i \leq k\}$ such that for all $(i, j) \notin \bar{R}$ the pair (V_i, V_j) is γ -regular and satisfies $d(V_i, V_j) = \eta_{i,j}$. The complexity of the regularity-instance is $\max(k, 1/\gamma)$.*

Note, that in the above definition the set \bar{R} corresponds to the set of pairs (i, j) for which (V_i, V_j) is not necessarily a γ -regular pair (possibly, there are at most $\gamma \binom{k}{2}$ such pairs). Also, note that the definition of a regularity-instance does not impose any restriction on the graphs spanned by any single set V_i . By Theorem 2.3, for any $0 < \gamma \leq 1$ any graph satisfies some regularity instance with an error parameter γ and with an order bounded by a function γ . The first step needed in order to obtain our characterization of the testable properties, is that the property of satisfying any given regularity-instance is testable. This is also the main technical result of this paper.

THEOREM 1. *For any regularity-instance R , the property of satisfying R is testable.*

2.1 The characterization

Many of the recent results on testing graph properties in the dense graph model relied on Lemma 2.3. Our main result in this paper shows that this is not a coincidence. Each of the papers which applied the regularity lemma to test a graph property used different aspects of what can be inferred from certain properties of a regular partition of a graph. These results however, use the properties of the regular partition in an *implicit* way. For example, the main observation needed in order to infer that triangle-freeness is testable, is that if the regular partition has three sets V_i, V_j, V_k , which are connected by regular and dense bipartite graphs, then the graph contains many triangles. However, to *test* triangle-freeness we do not need to know the regular partition, we just need to find a triangle in the graph. As Theorem 1 allows us to test for having a certain regular partition, it seems possible to try and test properties by *explicitly* checking for properties of the regular partition of the input. Returning to the previous discussion on viewing the regularity lemma as a constant complexity description of a graph, being able to explicitly test for having a given regular partition should allow us to test more complex properties as we can obtain all the information of the regular partition and not just *consequences* of having some regular partition. The next definition tries to capture the graph properties \mathcal{P} that can be tested via testing a certain set of regularity instances.

DEFINITION 2.5. (REGULAR-REDUCIBLE) *A graph property \mathcal{P} is regular-reducible if for any $\delta > 0$ there exists an $r = r(\delta)$ such that for any n there is a family \mathcal{R} of at most r regularity-instances each of complexity at most r , such that the following holds for every n -vertex graph G :*

1. *If G satisfies \mathcal{P} then for some $R \in \mathcal{R}$, G is δ -close to satisfying R .*
2. *If G is ϵ -far from satisfying \mathcal{P} , then for any $R \in \mathcal{R}$, G is $(\epsilon - \delta)$ -far from satisfying R .*

The reader may observe that in the above definition the value of δ may be arbitrarily close to 0. If we think of $\delta = 0$ then we get that a graph satisfies \mathcal{P} if and only if it satisfies one of the regularity instances of \mathcal{R} . With this interpretation in mind, in order to test \mathcal{P} one can test the property of satisfying any one of the instances of \mathcal{R} . Therefore, in some sense we “reduce” the testing of property \mathcal{P} to the testing of regularity-instances. As the main result of this paper states, the testable graph properties are precisely those for which testing them can be carried out by testing for some property of their regular partitions.

THEOREM 2. (MAIN RESULT) *A graph property is testable if and only if it is regular-reducible.*

If we have to summarize the moral of our characterization in one simple sentence, then it says that a graph property \mathcal{P} is testable if and only if \mathcal{P} is such that knowing a regular partition of a graph G is sufficient for telling whether G is ϵ -far or ϵ -close to satisfying \mathcal{P} . In other words, there is a short “proof” that G is either ϵ -close or ϵ -far from satisfying \mathcal{P} . Thus, in a more “computational complexity” jargon, we could say that a graph property is testable if and only if it has the following “interactive proof”: A prover gives a verifier the description of a regularity-instance R , which the input G is (supposedly) close to satisfying. The verifier, using Theorem 1, then verifies if G is indeed close to satisfying R . The way to turn this interactive proof into a testing algorithm is to apply the constant-complexity properties of the regularity lemma that we have previously discussed; as the order of the regular partition is bounded by a function of ϵ , there are only *finitely* many regularity-instances that the prover may potentially send to the verifier. Therefore, the verifier does not need to get an alleged regular-instances, it can simply try them all! Theorem 2 thus states that in some sense testing regularity-instances is the “hardest” property to test, because by Theorem 2 any testing algorithm can be turned into a testing algorithm for regularity-instances. However, we stress that this is true only on the *qualitative* level, because using Theorem 2 in order to turn a tester into a tester, which tests for regularity-instances may significantly increase its query complexity. The main reason is that the proofs of Theorems 1 and 2 apply Lemma 2.3 and thus only give weak upper bounds. Having said that, it should also be clear that one cannot prove general results on testing graph properties which guarantee good upper bounds (say, $\text{poly}(1/\epsilon)$) on the query complexity, as it was proved in [6] that there are graph properties (even monotone ones) that are testable and yet may require arbitrarily large query complexity. We also note that the terminology of regular-reducible is not far from being a standard reduction because in order to prove one of the directions of Theorem 2 we indeed test a property \mathcal{P} , which is regular-reducible to a set

\mathcal{R} , by testing the regularity-instances of \mathcal{R} . Theorem 2 also gives further convincing evidence to the “combinatorial” nature of property testing in the dense graph model as was recently advocated by Goldreich [23].

As is evident from Definition 2.5, the characterization given in Theorem 2 is not a “quick recipe” for inferring whether a given property is testable. Still, we can use Theorem 2 in order to obtain unified proofs for several previous results. As we have alluded to before, these results can be inferred by showing that it is possible (or impossible) to reduce the testing of the property to testing if a graph satisfies certain regularity-instances. We believe that these proofs give some (non-explicit) structural explanation as to what makes a graph property testable. See Section 7 for more details. It is thus natural to ask if one can come up with more “handy” characterizations. We doubt that such a characterization exists, mainly because it should (obviously) be equivalent to Theorem 2. Of course, we cannot formally prove that no simpler sufficient condition exists. However, as we discuss in the next subsection we can at least disprove a possible simpler sufficient condition of testability.

2.2 Disproving a simpler sufficient condition

Observing previous results in property testing (not necessarily of graphs) reveals that essentially all the properties that were shown to be testable had the following *downscaling* property: If an object is close to satisfying a property then a sample from the object will not be very far from satisfying it. This is true, for example, for properties studied in the context of functions [16, 19], graphs [24, 7, 13], geometric objects [2, 11], algebraic structures [10, 20, 8] and languages [4, 31]. Therefore, a natural question is whether being downscaling is enough for being testable. To formally state this feature for graphs we introduce the following definition:

DEFINITION 2.6. (DOWNSCALING) *A graph property \mathcal{P} is downscaling if for every $\delta > 0$ there is a $q = q(\delta)$ with the following property: suppose that an n -vertex graph is ϵ -close to satisfying \mathcal{P} . Then, for every m such that $q \leq m \leq n$ the graph induced by a randomly chosen set of vertices of size m is $(\epsilon + \delta)$ -close to satisfying \mathcal{P} with probability at least $2/3$.*

We note that essentially the same definition as above applies to other combinatorial structures. The reader may want to note that the family of hereditary properties, which were shown to be testable in [7] (see also [28]), are all downscaling. Also, all the partition properties, which were shown to be testable in [24], are downscaling. If downscaling was indeed sufficient for being testable, this would immediately give simple and uniform proofs for many testability results. As it turns out however, this is not the case.

THEOREM 3. *There exists a downscaling non-testable graph property.*

2.3 Organization and overview of the paper

The first main technical step of the proof of Theorem 1 is taken in Section 3. In this section we prove that if the densities of pairs of subsets of vertices of a bipartite graph are close to the density of the bipartite graph itself, then the bipartite graph can be turned into a regular-pair using relatively few edge modifications. Rephrasing this gives that we can increase the regularity measure of a bipartite

pair by making relatively few edge modifications. The second main step is taken in Section 5. In this section, we show that sampling a constant number of vertices guarantees that the sample and the graph will have (roughly) the *same* set of regular partitions. We believe that this result may be of independent interest. By applying the results of Sections 3 and 5 we prove Theorem 1 in Section 6. In this section we also prove one of the directions of Theorem 2, asserting that if a graph property is regular-reducible then it is testable. Along with Theorem 1, a second tool that we need in order to prove this direction is the main result of [17]. We apply this result in order to infer that for any regularity-instance R , one can not only test the property of satisfying R , but can also estimate how far is a given graph from satisfying R . This *estimation* of the distance to satisfying regularity-instances is key to *testing* a property via a regularity-reduction. The proof of the second direction of Theorem 2 appears in Section 4. To prove this direction we first show that knowing that a graph G satisfies a regularity instance enables us to estimate the number of copies of certain graphs in G . We then apply the main result of [25] about canonical testers along with the main result of Section 3 in order to “pick” those regularity-instances that can constitute the family \mathcal{R} in Definition 2.5. In Section 7 we use Theorem 2 in order to reprove some previously known results in property-testing. The main interest of these proofs is that they apply Theorem 2 in order to prove in a unified manner results that had distinct proofs. Due to space limitations, the proof of Theorem 3 will appear only in the full version of the paper. We briefly mention that in order to prove this theorem we use a subtle variation of the graph isomorphism problem, which was known to be non-testable but is far from being downscaling. Section 8 contains some concluding remarks.

3. ENHANCING REGULARITY WITH FEW EDGE MODIFICATIONS

The definition of a γ -regular pair of density η requires a pair of sets of vertices to satisfy several density requirements. Our main goal in this section is to show that if a pair of vertex sets are close (in an appropriate sense) to satisfying these requirements, then it is indeed close to being a γ -regular pair of density η . For example, consider the property of being a 0.1-regular pair with edge density 0.5. Intuitively, it seems that if the edge density of a bipartite graph G on vertex sets A and B of size m each is close to 0.5, and the density of any pair $A' \subseteq A$ and $B' \subseteq B$ of sizes $0.1m$ is close to 0.5 ± 0.1 , then G should be close to satisfying the property. However, note that it may be the case that there are pairs (A', B') , whose density is smaller than 0.4, and other pairs, whose density is larger than 0.6. Thus, only removing or only adding edges (even randomly) will most likely not turn G into a 0.1-regular pair of density 0.5. In order to show that G is indeed close to satisfying the property, we take a “convex combination” of G with a random graph, whose density is $1/2$. The intuition is that the random graph will not change the density of G much, but, because a random graph is highly regular, it will increase the regularity of G . The main result of this section is formalized in the following lemma, which is an important ingredient in the proofs of both directions of Theorem 2.

In this lemma, as well as throughout the rest of the paper,

when we write $x = a \pm b$ we mean $a - b \leq x \leq a + b$.

LEMMA 3.1. *The following holds for any $0 < \delta \leq \gamma \leq 1$: Suppose that (A, B) is a $(\gamma + \delta)$ -regular pair with density $\eta \pm \delta$, where $|A| = |B| = m \geq m_{3.1}(\eta, \delta)$. Then, it is possible to make at most $50 \frac{\delta}{\gamma^2} m^2$ edge modifications and turn (A, B) into a γ -regular pair with density precisely η .*

The proof of Lemma 3.1 has two main steps, which are captured in Lemmas 3.2 and 3.3 below. The proofs will appear in the full version of the paper. The first step is given in the following lemma, which enables us to make relatively few edge modifications and thus make sure that the density of a pair is exactly what it should be, while at the same time not decreasing its regularity by much.

LEMMA 3.2. *Suppose that (A, B) is a $(\gamma + \delta)$ -regular pair satisfying $d(A, B) = \eta \pm \delta$, where $|A| = |B| = m \geq m_{3.2}(\eta, \delta)$. Then, it is possible to make at most $2\delta m^2$ modifications, and thus turn (A, B) into a $(\gamma + 2\delta)$ -regular pair with density precisely η .*

The second and main step, which implements the main idea presented at the beginning of this section, takes a bipartite graph, whose density is precisely η , and returns a bipartite graph, whose density is still η but with a better regularity measure.

LEMMA 3.3. *The following holds for any $0 < \delta \leq \gamma \leq 1$. Let A and B be two vertex sets of size $m \geq m_{3.3}(\delta, \gamma)$, satisfying $d(A, B) = \eta$. Suppose further that for any pair of subsets $A' \subseteq A$ and $B' \subseteq B$ of size γm we have $d(A', B') = \eta \pm (\gamma + \delta)$. Then, it is possible to make at most $\frac{3\delta}{\gamma} m^2$ edge modifications and thus turn (A, B) into a γ -regular pair with density precisely η .*

The following application of Lemma 3.1 will be useful later in the paper.

LEMMA 3.4. *Let R be a regularity-instance of order k , error-parameter γ , $\binom{k}{2}$ edge densities $\eta_{i,j}$ and set of non-regular pairs \bar{R} . If a graph G has an equipartition $\mathcal{V} = \{V_1, \dots, V_k\}$ of order k such that*

1. $d(V_i, V_j) = \eta_{i,j} \pm \frac{\gamma^2 \epsilon}{50}$ for all $i < j$.
2. Whenever $(i, j) \notin \bar{R}$, the pair (V_i, V_j) is $(\gamma + \frac{\gamma^2 \epsilon}{50})$ -regular.

Then G is ϵ -close to satisfying R .

PROOF. For any $(i, j) \notin \bar{R}$ we can use Lemma 3.1 and make at most $50 \frac{\gamma^2 \epsilon / 50}{\gamma^2} (n/k)^2 \leq \epsilon n^2 / k^2$ edge modifications to turn (V_i, V_j) into a γ -regular pair with density $\eta_{i,j}$. As there are at most $\binom{k}{2}$ pairs this is a total of at most ϵn^2 modifications. We have thus turned G into a graph satisfying R by making at most ϵn^2 edge modifications, as needed. \square

4. A TESTABLE PROPERTY IS REGULAR-REDUCIBLE

In this section we prove the first direction of Theorem 2.

LEMMA 4.1. *If a graph property is testable then it is regular-reducible.*

Our starting point in the proof of Lemma 4.1 is the following result of [25] (extending a result of [3]) about canonical testers:

LEMMA 4.2 ([3, 25]). *If a graph property \mathcal{P} can be tested on n -vertex graphs with $q = q(\epsilon, n)$ edge queries, then it can also be tested by a tester, which makes its queries by uniformly and randomly choosing a set of $2q$ vertices, querying all the pairs and then accepting or rejecting (deterministically) according to the graph induced by the sample, the value of ϵ and the value of n .*

Restating the above, by (at most) squaring the query complexity, we can assume without loss of generality that a property-tester works by sampling a set of vertices of size $q(\epsilon, n)$ and accepting or rejecting according to some graph property of the sample. As noted in [25], the graph property that the algorithm may search for in the sample may be different from the property, which is tested. In fact, the property the algorithm checks for in the sample may depend on ϵ and on the size of the input graph. Our main usage of Lemma 4.2 is that it allows to pick the graphs of size q that cause a tester for \mathcal{P} to accept. The first technical step that we take towards proving Lemma 4.1 is proving some technical results about induced copies of graphs spanned by graphs satisfying a given regularity-instance. These results enable us to deduce from the fact that a graph satisfies some regularity-instance the probability that a given tester accepts the graph. We then use these results along with Lemma 4.2 and some additional arguments in order to prove that any testable property is regular reducible. The details follow.

DEFINITION 4.3. *Let H be a graph on h vertices, let W be a weighted complete graph on h vertices, where the weight of an edge (i, j) is $\eta_{i,j}$. For a permutation $\sigma : [h] \rightarrow [h]$ define*

$$IC(H, W, \sigma) = \prod_{(i,j) \in E(H)} \eta_{\sigma(i), \sigma(j)} \prod_{(i,j) \notin E(H)} (1 - \eta_{\sigma(i), \sigma(j)})$$

Suppose V_1, \dots, V_k are k vertex sets, each of size m , and suppose the bipartite graph spanned by V_i and V_j is a bipartite random graph with edge density $\eta_{i,j}$. Let H be a graph of size k , and let $\sigma : [k] \rightarrow [k]$ be some permutation. What is the expected number of k -tuples of vertices $v_1 \in V_1, \dots, v_k \in V_k$, which span an induced copy of H with each v_i playing the role of $\sigma(i)$? It is easy to see that the answer is $IC(H, W, \sigma) m^k$, where W is the weighted complete graph with weights $\eta_{i,j}$. The following claim shows that this is approximately the case when instead of random bipartite graphs we take regular enough bipartite graphs. The proof is a standard application of the definition of a regular pair and is thus omitted from this extended abstract. See Lemma 4.2 in [15] for a version of the proof.

CLAIM 4.4. *For any δ and h , there exists a $\gamma = \gamma_{4.4}(\delta, h)$ such that the following holds: Suppose V_1, \dots, V_h are h sets of vertices of size m each, and that all the pairs (V_i, V_j) are γ -regular. Define W to be the weighted complete graph on h vertices, whose weights are $\eta_{i,j} = d(V_i, V_j)$. Then, for any graph H on h vertices and for any $\sigma : [h] \rightarrow [h]$, the number of h -tuples $v_1 \in V_1, \dots, v_h \in V_h$, which span an induced copy of H with each v_i playing the role of the vertex $\sigma(i)$ is $(IC(H, W, \sigma) \pm \delta) m^h$*

We would now want to consider the total number of induced copies of some graph.

DEFINITION 4.5. Let H be a graph on h vertices, let W be a weighted complete graph on h vertices, where the weight of edge (i, j) is $\eta_{i,j}$. Let $\text{Aut}(H)$ denote the number of automorphisms of H . Define

$$IC(H, W) = \frac{1}{\text{Aut}(H)} \sum_{\sigma} IC(H, W, \sigma).$$

Continuing the discussion before Claim 4.4, it is easy to see that in this case the expected number of induced copies of H having one vertex in each of the sets V_i is $IC(H, W)$. Again, we can show that the same is approximately true when we replace random bipartite graphs with regular enough bipartite graphs.

CLAIM 4.6. For any δ and k , there exists a $\gamma = \gamma_{4.6}(\delta, k)$ such that the following holds: Suppose that V_1, \dots, V_k are sets of vertices of size m each, and that all the pairs (V_i, V_j) are γ -regular. Define K to be the weighted complete graph on k vertices, whose weights are $\eta_{i,j} = d(V_i, V_j)$. Then, for any graph H of size k , the number of induced copies of H , which have precisely one vertex in each of the sets V_1, \dots, V_k is $(IC(H, W) \pm \delta)m^k$.

PROOF. Set $\gamma_{4.6}(\delta, k) = \gamma_{4.4}(\delta/k!, k)$. Suppose V_1, \dots, V_k are as in the statement of the claim and let H be any graph on k vertices. By Claim 4.4 for every permutation $\sigma : [k] \rightarrow [k]$, the number of induced copies of H which have precisely one vertex v_i in each set V_i such that v_i plays the role of vertex $\sigma(i)$ is $IC(H, W, \sigma) \pm \delta m^k/k!$. If we sum over all permutations $\sigma : [k] \rightarrow [k]$ we get $\sum_{\sigma} (IC(H, W, \sigma) \pm \delta/k!)m^k$. This summation, however, counts copies of H several times. More precisely, each copy is thus counted $\text{Aut}(H)$ times. Thus, dividing by $\text{Aut}(H)$ gives that the number of such induced copies is

$$\frac{1}{\text{Aut}(H)} \left(\sum_{\sigma} (IC(H, W, \sigma) \pm \delta/k!)m^k \right) =$$

$$\left(\frac{1}{\text{Aut}(H)} \sum_{\sigma} IC(H, W, \sigma) \pm \delta \right) m^k = (IC(H, W) \pm \delta)m^k.$$

□

We would now want to consider the number of induced copies of a graph H , when the number of sets V_i is larger than the size of H .

DEFINITION 4.7. Let H be a graph on h vertices, let R be a weighted complete graph of size at least h where the weight of an edge (i, j) is $\eta_{i,j}$, and let \mathcal{W} denote all the subsets of $V(W)$ of size h . Define

$$IC(H, R) = \sum_{W \in \mathcal{W}} IC(H, W).$$

The following lemma shows that knowing that a graph satisfies some regularity-instance R , enables us to estimate the number of induced copies spanned by any graph, which satisfies R .

LEMMA 4.8. For any δ and q , there are $k = k_{4.8}(\delta, q)$ and $\gamma = \gamma_{4.8}(\delta, q)$ with the following properties: For any regularity-instance R of order at least k and with error parameter at most γ , and for every graph H of size $h \leq q$, the number of induced copies of H in any n -vertex graph satisfying R is $(IC(H, R) \pm \delta) \binom{n}{h}$.

PROOF. Put $k = k_{4.8}(\delta, q) = \frac{\delta}{10q^2}$ and $\gamma = \gamma_{4.8}(\delta, q) = \min\{\frac{\delta}{3q^2}, \gamma_{4.6}(\frac{1}{3}\delta, q)\}$. Let R be any regularity instance as in the statement, let G be any graph satisfying R , and let H be any graph of size $h \leq q$. Let V_1, \dots, V_{ℓ} be an equipartition of G satisfying R . For the proof of the lemma it will be simpler to consider an equivalent statement of the lemma, stating that if one samples an h -tuple of vertices from G , then the probability that it spans an induced copy of H is $IC(H, R) \pm \delta$.

First, note that by our choice of k we get from a simple birthday-paradox argument, that the probability that the h -tuple of vertices has more than one vertex in any one of the sets V_i is at most $\frac{1}{3}\delta$. Second, observe that as the equipartition of R is γ -regular and $\gamma \leq \delta$, we get that the probability that the h -tuple of vertices contains a pair $v_i \in V_i$ and $v_j \in V_j$ such that (V_i, V_j) is not γ -regular is at most $\binom{h}{2}\gamma \leq \binom{q}{2}\gamma \leq \frac{1}{3}\delta$. Thus, it is enough to show that conditioning on the events: (i) the h vertices v_1, \dots, v_h belong to distinct sets V_i , (ii) if $v_i \in V_i, v_j \in V_j$ and (V_i, V_j) is γ -regular, then the probability that they span an induced copy of H is $IC(H, R) \pm \frac{1}{3}\delta$. Assuming events (i) and (ii) hold let us compute the probability that the h -tuple of vertices spans an induced copy of H , while conditioning on the h sets from V_1, \dots, V_{ℓ} which contain the h vertices. For every possible set W of h sets V_i we get from the choice of γ via Claim 4.6 that the probability that they span an induced copy of H is $IC(H, W) \pm \frac{1}{3}\delta$. This means that the conditional probability that the h -tuple of vertices span an induced copy of H is $IC(H, R) \pm \frac{1}{3}\delta$, as needed. □

PROOF. (OF LEMMA 4.1): Suppose \mathcal{P} is testable by a tester T , and assume without loss of generality that T is canonical. This assumption is possible by Lemma 4.2. Let $q(\epsilon)$ be the upper bound guarantee for the query complexity of T . Fix any n and δ and assume that $\delta < 1/12$ (otherwise, replace δ with $1/13$). Let $q = q(\delta, n) \leq q(\delta)$ be the query complexity, which is sufficient for T to distinguish between n -vertex graphs satisfying \mathcal{P} and those that are δ -far from satisfying it, with success probability at least $2/3$. As T is canonical, if it samples a set of vertices and gets a graph of size q , it either rejects or accepts deterministically. Hence, we can define a set \mathcal{A} , of all the graphs Q of size q , such that if the sample of vertices spans a graph isomorphic to Q , then T accepts the input. We finally put $k = k_{4.8}(\delta/2^{\binom{2}{2}}, q)$, $\gamma = \gamma_{4.8}(\delta/2^{\binom{2}{2}}, q)$ and $T = T_{2.3}(k, \gamma)$. For any $k \leq t \leq T$ consider all the (finitely many) regularity-instances of order t , where for the edge densities $\eta_{i,j}$ we choose a real from the set $\{0, \frac{\delta\gamma^2}{50q^2}, 2\frac{\delta\gamma^2}{50q^2}, 3\frac{\delta\gamma^2}{50q^2}, \dots, 1\}$. Let \mathcal{I} be the union of all these regularity-instances. Note, that all the above constants, as well as the size of \mathcal{I} and the complexity of the regularity-instances in \mathcal{I} , are determined as a function of δ only (and the property \mathcal{P}).

We claim that we can take \mathcal{R} in Definition 2.5 to be

$$\mathcal{R} = \{R \in \mathcal{I} : \sum_{H \in \mathcal{A}} IC(R, H) \geq 1/2\}.$$

To see this, first note that the expression $\sum_{H \in \mathcal{A}} IC(R, H)$ is an estimation of the fraction of induced copies of graphs from \mathcal{A} in a graph satisfying R . Combining the facts that the graphs in \mathcal{A} all have size q and the use of Lemma 4.8 with $\delta/2 \binom{q}{2}$ we infer that the expression $\sum_{H \in \mathcal{A}} IC(R, H)$ is an estimate of the number of induced copies of graphs from \mathcal{A} in a graph satisfying R , up to an additive error of at most $\delta \binom{n}{q}$.

Suppose a graph G satisfies \mathcal{P} . This means that T accepts G with probability at least $2/3$. In other words, this means that at least $\frac{2}{3} \binom{n}{q}$ of the subsets of q vertices of G span a graph isomorphic to one of the members of \mathcal{A} . By Lemma 2.3 G has some γ -regular partition of size at least k and at most T . As the densities in the regularity-instances in \mathcal{A} differ by $\frac{\delta \gamma^2}{50q^2}$ we get that the densities of the regular partition of G differ by at most $\frac{\delta \gamma^2}{50q^2}$ from the densities of one of the regularity-instances $R \in \mathcal{I}$. Lemma 3.4 implies that G is δ/q^2 -close to satisfying one of the regularity-instances of \mathcal{I} . Note that adding and/or removing an edge can decrease the number of induced copies of members of \mathcal{A} in G by at most $\binom{n-2}{q-2}$. Thus adding and/or removing $\delta n^2/q^2$ edges can decrease the number of induced copies of members of \mathcal{A} in G by at most $\delta \frac{n^2}{q^2} \binom{n-2}{q-2} \leq \delta \binom{n}{q}$. Thus, after these at most $\delta n^2/q^2$ edge modifications we get a graph that satisfies one of the regularity-instances $R \in \mathcal{I}$ where at least $(\frac{2}{3} - \delta) \binom{n}{q} > (\frac{1}{2} + \delta) \binom{n}{q}$ of the subsets of q vertices of the new graph span a member of \mathcal{A} (here we use the assumption that $\delta < 1/12$). As explained in the previous paragraph, by our choice of k and γ via Lemma 4.8, this means that $\sum_{H \in \mathcal{A}} IC(R, H) \geq 1/2$. By the definition of \mathcal{R} this means that $R \in \mathcal{R}$, so G is indeed δ -close to satisfying one of the regularity-instances of \mathcal{R} .

Suppose now that a graph G is ϵ -far from satisfying \mathcal{P} . If $\delta \geq \epsilon$ then there is nothing to prove, so assume that $\delta < \epsilon$. If G is $(\epsilon - \delta)$ -close to satisfying a regularity-instance $R \in \mathcal{R}$, then by the definition of \mathcal{R} and our choice of k and γ via Lemma 4.8 it is $(\epsilon - \delta)$ -close to a graph G' , such that at least $(\frac{1}{2} - \delta) \binom{n}{q} > (\frac{1}{3} + \delta) \binom{n}{q}$ of the subsets of q vertices of G' span an induced copy of a graph from \mathcal{A} . In other words, this means that T accepts G' with probability at least $\frac{1}{3} + \delta$. This means that G' cannot be δ -far from satisfying \mathcal{P} as we assume that q is enough for T to reject with probability at least $2/3$ graphs that are δ -far from satisfying \mathcal{P} . However, as G is ϵ -far from satisfying \mathcal{P} any graph that is $(\epsilon - \delta)$ -close to G must be δ -far from satisfying \mathcal{P} , a contradiction. \square

5. SAMPLING REGULAR PARTITIONS

The main result of this section (roughly) asserts that for every fixed γ , if we sample a constant number of vertices from a graph G , then with high probability the graph induced by the sample and the graph G will have the same set of γ -regular partitions. The proofs of this section will appear in the full version of the paper. To formally state this result we introduce the following definition:

DEFINITION 5.1. (δ -SIMILAR REGULAR-PARTITION) *We say that an equipartition $\mathcal{U} = \{U_i \mid 1 \leq i \leq k\}$ is δ -similar to a γ -regular equipartition $\mathcal{V} = \{V_i \mid 1 \leq i \leq k\}$, of the same order k (where $0 < \gamma \leq 1$), if: (1) $d(U_i, U_j) = d(V_i, V_j) \pm \delta$ for all $i < j$. (2) Whenever (V_i, V_j) is γ -regular, (U_i, U_j) is $(\gamma + \delta)$ -regular.*

Observe that in the above definition, the two equipartitions \mathcal{V} and \mathcal{U} may be equipartitions of different graphs. In what follows, if $G = (V, E)$ is a graph and $Q \subseteq V(G)$, then $G[Q]$ denotes the subgraph induced by G on Q . Our main result in this section is the following:

LEMMA 5.2. *For every k, δ there exists $q = q_{5.2}(k, \delta)$ such that a sample Q , of q vertices from a graph G , satisfies the following with probability at least $2/3$: If G has a γ -regular equipartition \mathcal{V} of order at most k , then $G[Q]$ has an equipartition \mathcal{U} , which is δ -similar to \mathcal{V} . Also, If $G[Q]$ has a γ -regular equipartition \mathcal{U} of order at most k , then G has an equipartition \mathcal{V} , which is δ -similar to \mathcal{U} .*

The proof of Lemma 5.2 has two main stages. For the first one we need a weaker result, which says that a sample of vertices has a regular partition, but with a *weaker* regularity measure.

LEMMA 5.3 ([13]). *For every k and γ there exists $q = q_{5.3}(k, \gamma)$ such that if a graph G has a γ -regular equipartition $\mathcal{V} = \{V_1, \dots, V_k\}$ of order k , then with probability at least $2/3$, a sample of q vertices will have an equipartition $\mathcal{U} = \{U_1, \dots, U_k\}$ satisfying: (1) $d(U_i, U_j) = d(V_i, V_j) \pm \delta$ for all $i < j$. (2) Whenever (V_i, V_j) is γ -regular (U_i, U_j) is $50\gamma^{1/5}$ -regular.*

For our purposes however, we cannot allow a weaker regularity as in the above lemma. Our main tool in the proof of Lemma 5.2 is Lemma 5.5 below, which establishes that if two graphs share *one* γ -regular equipartition, then they share *all* the γ' -regular-partitions where γ' is slightly larger than γ . This will allow us to strengthen Lemma 5.3 and thus obtain Lemma 5.2. For the statement of this lemma we need the following definition:

DEFINITION 5.4. ((δ, γ) -SIMILAR REGULAR-PARTITIONS) *Two equipartitions $\mathcal{V} = \{V_i \mid 1 \leq i \leq k\}$ and $\mathcal{U} = \{U_i \mid 1 \leq i \leq k\}$ of the same order k , are said to be (δ, γ) -similar if: (1) $d(U_i, U_j) = d(V_i, V_j) \pm \delta$ for all $i < j$. (2) For all but at most $\gamma \binom{k}{2}$ of the pairs $i < j$, both (V_i, V_j) and (U_i, U_j) are γ -regular.*

LEMMA 5.5. *For every k and δ there exists $\zeta = \zeta_{5.5}(k, \delta)$ with the following property: suppose that two graphs $G = (V, E)$ and $\bar{G} = (\bar{V}, \bar{E})$ have (ζ, ζ) -similar regular equipartitions $\mathcal{V} = \{V_1, \dots, V_\ell\}$ and $\bar{\mathcal{V}} = \{\bar{V}_1, \dots, \bar{V}_\ell\}$ with $\ell \geq 1/\zeta$. Then, if \bar{G} has a γ -regular equipartition $\bar{\mathcal{A}} = \{\bar{A}_1, \dots, \bar{A}_k\}$ then G has an equipartition $\mathcal{A} = \{A_1, \dots, A_k\}$, which is δ -similar to $\bar{\mathcal{A}}$.*

6. TESTING REGULAR PARTITIONS AND PROOF OF THE MAIN RESULT

In this section we apply the results of Sections 3 and 5 to prove Theorem 2. We start by proving the main technical result of this paper by showing that the property of satisfying a regularity-instance is testable with a constant number of queries.

PROOF. (OF THEOREM 1): Suppose the regularity-instance R has error parameter γ , $\binom{k}{2}$ edge densities $\eta_{i,j}$, and a set of non-regular pairs \bar{R} . Given $G = (V, E)$, the algorithm for testing the property of having R , samples a set of vertices Q , of size q , where q will be chosen later, and accepts G if and

only if the graph spanned by Q is $\frac{\gamma^4 \epsilon}{200k^2}$ -close to satisfying R . In what follows we denote by $G[Q]$ the graph spanned by Q .

CLAIM 1: If G satisfies R , and $q \geq q_1(\epsilon, k, \gamma)$, then $G[Q]$ is $\frac{\gamma^4 \epsilon}{200k^2}$ -close to satisfying R with probability at least $2/3$.

PROOF. If $G = (V, E)$ satisfies R , then V has an equipartition into V_1, \dots, V_k such that for all $(i, j) \notin \bar{R}$ the pair (V_i, V_j) is γ -regular. If we take $q_1(\epsilon, k, \gamma) = q_{5.2}(k, \frac{\gamma^6 \epsilon}{10000k^2})$, then by Lemma 5.2, with probability at least $2/3$ the graph $G[Q]$ will have an equipartition into k sets A_1, \dots, A_k , such that $d(A_i, A_j) = \eta_{i,j} \pm \frac{\gamma^6 \epsilon}{10000k^2}$ for all $i < j$, and if (V_i, V_j) is γ -regular then (A_i, A_j) is $(\gamma + \frac{\gamma^6 \epsilon}{10000k^2})$ -regular. By Lemma 3.4, this means that $G[Q]$ is $\frac{\gamma^4 \epsilon}{200k^2}$ -close to satisfying R . \square

CLAIM 2: If G is ϵ -far from satisfying R , and $q \geq q_2(\epsilon, k, \gamma)$, then $G[Q]$ is $\frac{\gamma^4 \epsilon}{200k^2}$ -far from satisfying R with probability at least $2/3$.

PROOF. We take $q_2(\epsilon, k, \delta) = q_{5.2}(k, \frac{\gamma^4 \epsilon}{200k^2})$. By Lemma 5.2 we get that with probability at least $2/3$ the graph $G[Q]$ is such that if it has a γ' -regular equipartition of order k , then G has an equipartition which is $\frac{\gamma^4 \epsilon}{200k^2}$ -similar to it. We claim that if this event occurs then $G[Q]$ is $\frac{\gamma^4 \epsilon}{200k^2}$ -far from satisfying R , which is what we want to show. Suppose $G[Q]$ satisfies the above property and assume that none the less it is $\frac{\gamma^4 \epsilon}{200k^2}$ -close to satisfying R . Consider the $\frac{\gamma^4 \epsilon}{200k^2} q^2$ edge modifications that make $G[Q]$ satisfy R and consider an equipartition $\mathcal{U} = \{U_1, \dots, U_k\}$ of $G[Q]$, which satisfies R after performing these modifications. As we made at most $\frac{\gamma^4 \epsilon}{200k^2} q^2$ edge modifications, we initially had $d(U_i, U_j) = \eta_{i,j} \pm \frac{\gamma^4 \epsilon}{200}$. Consider now any $(i, j) \notin \bar{R}$. After these modifications (U_i, U_j) must be γ -regular with density $\eta_{i,j}$. Therefore, after these modifications every pair $U'_i \subseteq U_i, U'_j \subseteq U_j$ satisfying $|U'_i| \geq \gamma|U_i|$ and $|U'_j| \geq \gamma|U_j|$ satisfies $d(U'_i, U'_j) = \eta_{i,j} \pm \gamma$. Hence, before the modifications every such pair satisfied $d(U'_i, U'_j) = \eta_{i,j} \pm (\gamma + \frac{\gamma^2 \epsilon}{200})$. Note that this means that every such pair was originally $(\gamma + \frac{\gamma^2 \epsilon}{100})$ -regular. By our assumption on $G[Q]$ this means that G has an equipartition in V_1, \dots, V_k such that $d(V_i, V_j) = \eta_{i,j} \pm \frac{\gamma^2 \epsilon}{50}$ holds for all $i < j$, and for all $(i, j) \notin \bar{R}$ the pair (V_i, V_j) is $(\epsilon + \frac{\gamma^2 \epsilon}{50})$ -regular. By Lemma 3.4, this means that G is ϵ -close to satisfying R , contradicting our assumption. \square

Combining the above two claims we infer that if $q \geq \max(q_1(\epsilon, k, \gamma), q_2(\epsilon, k, \gamma))$ then with probability at least $2/3$ the algorithm distinguishes between the required two cases. Furthermore, the number of queries performed by the algorithm depends only on ϵ, k and γ , and is thus bounded from above by a function of ϵ and r . This completes the proof of the theorem. \square

Having established the testability of any given regularity-instance we can prove Theorem 2. The last tool we need for the proof is the main result of [17] about estimating graph properties.

THEOREM 4 ([17]). *Suppose that a graph property \mathcal{P} is testable. Then for every $0 \leq \epsilon_1 < \epsilon_2 \leq 1$ there is a randomized algorithm for distinguishing between graphs that are*

ϵ_1 -close to satisfying \mathcal{P} and graphs that are ϵ_2 -far from satisfying it. Furthermore, the query complexity of the algorithm can be bounded from above by a function of ϵ_1 and ϵ_2 , which is independent of the size of the input.

PROOF. (OF THEOREM 2): The first direction is given in Lemma 4.1. For the other direction, suppose that a graph property \mathcal{P} is regular-reducible as per Definition 2.5. Let us fix n and ϵ . Put $r = r(\frac{1}{4}\epsilon)$ and let \mathcal{R} be the corresponding set of regularity instances for $\delta = \frac{1}{4}\epsilon$ as in Definition 2.5. Recall that Definition 2.5 guarantees that the number and the complexity of the regularity-instances of \mathcal{R} are bounded by a function of $\delta = \frac{1}{4}\epsilon$. By Theorem 1 for any regularity-instance $R \in \mathcal{R}$, the property of satisfying R is testable. Thus, by Theorem 4 for any such R , we can distinguish graphs that are $\frac{1}{4}\epsilon$ -close to satisfying R from those that are $\frac{3}{4}\epsilon$ -far from satisfying it, while making a number of queries, which is bounded by a function of ϵ . In particular, by repeating the algorithm of Theorem 4 an appropriate number of times (that depends only on $r = r(\frac{1}{4}\epsilon)$), and taking the majority vote, we get an algorithm for distinguishing between the above two cases, whose query complexity is a function of ϵ and r , which succeeds with probability at least $1 - \frac{1}{3r}$. As r itself is bounded by a function of ϵ , the number of queries of this algorithm can be bounded by a function of ϵ only.

We are now ready to describe our tester for \mathcal{P} : Given a graph G of size n and $\epsilon > 0$, the algorithm uses for every $R \in \mathcal{R}$ the version of Theorem 4 described in the previous paragraph, which succeeds with probability at least $1 - \frac{1}{3r}$ in distinguishing between that case that G is $\frac{1}{4}\epsilon$ -close to satisfying R from the case that it is $\frac{3}{4}\epsilon$ -far from satisfying it. If it finds that G is $\frac{1}{4}\epsilon$ -close to satisfying some $R \in \mathcal{R}$, then the algorithm accepts, and otherwise it rejects. Observe that as there are at most r regularity-instances in \mathcal{R} , we get by the union-bound that with probability at least $2/3$ the subroutine for estimating how far is G from satisfying some $R \in \mathcal{R}$ never errs. We now prove that the above algorithm is indeed a tester for \mathcal{P} . Suppose first that G satisfies \mathcal{P} . As we set $\delta = \frac{1}{4}\epsilon$ and \mathcal{P} is regular-reducible to \mathcal{R} , the graph G must be $\frac{1}{4}\epsilon$ -close to satisfying some regularity-instance $R' \in \mathcal{R}$. Suppose now that G is ϵ -far from satisfying \mathcal{P} . Again, as we assume that \mathcal{P} is regular-reducible to \mathcal{R} , we conclude that G must be $\frac{3}{4}\epsilon$ -far from satisfying all of the regularity-instances $R \in \mathcal{R}$. As with probability at least $2/3$ the algorithm correctly decides for any $R \in \mathcal{R}$ if G is $\frac{1}{4}\epsilon$ -close to satisfying R or $\frac{3}{4}\epsilon$ -far from satisfying it, we get that if G satisfies \mathcal{P} then with probability at least $2/3$ the algorithm will find that G is $\frac{1}{4}\epsilon$ -close to satisfying some $R \in \mathcal{R}$, while if G is ϵ -far from satisfying \mathcal{P} then with probability at least $2/3$ the algorithm will find that G is $\frac{3}{4}\epsilon$ -far from all $R \in \mathcal{R}$. By the definition of the algorithm, we get that with probability at least $2/3$ it distinguishes between graphs satisfying \mathcal{P} from those that are ϵ -far from satisfying it. This means that the algorithm is indeed a tester for \mathcal{P} . \square

7. APPLICATIONS OF THE MAIN RESULT

In this section we show that Theorem 2 can be used in order to derive some positive and negative results on testing graph properties. We would like to stress that all these proofs implicitly apply the main intuition behind our characterization, which was explained after the statement of Theorem 2, that a graph property is testable if and only

if knowing the regularity partition of the graph is sufficient for inferring if a graph is far from satisfying the property. Our first application of Theorem 2 concerns testing for H -freeness; A graph is said to be H -free if it contains no (not necessarily induced) copy of H . It was implicitly proved in [1] that for any H , the property of being H -free is testable. The main idea of the proof in [1] is that if G is ϵ -far from being H -free then a large enough sample of vertices will contain a copy of H with high probability. Here we derive this result from Theorem 2 by giving an alternative proof, which checks if the input satisfies some regularity-instance. For simplicity, we only consider testing triangle-freeness. We briefly mention that an argument similar to the one we use to test triangle-freeness can be used to test any monotone graph property. However, to carry out the proof one needs one additional non-trivial argument, which was proved in [6], so we refrain from including the proof.

COROLLARY 7.1. *Triangle-freeness is testable.*

PROOF. (SKETCH): By Theorem 2 it is enough to show that triangle-freeness is regular-reducible. Fix any $\delta > 0$ and set $\gamma' = \gamma_{4.6}(\delta, 3)$. Define $\gamma = \min\{\gamma', \delta\}$. We define \mathcal{R} to be all the regularity-instances R satisfying the following: (i) They have regularity parameter γ (ii) They have order at least $1/\gamma$ and at most $T_{2.3}(1/\gamma, \gamma)$ (iii) Their densities $\eta_{i,j}$ are taken from $\{0, \gamma, 2\gamma, \dots, 1\}$. (iv) They do not contain three clusters V_i, V_j, V_k such that $\eta_{i,j}, \eta_{j,k}, \eta_{i,k}$ are all positive.

To show that this is a valid reduction, assume first that G is ϵ -far from being triangle-free. Assume G is $(\epsilon - \delta)$ -close to satisfying a regularity instance $R \in \mathcal{R}$. We can thus make $(\epsilon - \delta)n^2$ edge modifications and get a graph satisfying R . We also remove all edges inside the sets V_i . As by item (ii) each set has size at most $\gamma n \leq \delta n$ we remove less than δn^2 edges. The total number of edges removed is thus less than ϵn^2 . By property (iv) of the regularity instances of \mathcal{R} this means that the new graph is triangle-free, which is impossible because we made less than ϵn^2 edge modifications and G was assumed to be ϵ -far from being triangle-free. Assume now that G is triangle-free. By Lemma 2.3 G has a γ -regular equipartition V_1, \dots, V_k of order $1/\gamma \leq k \leq T_{2.3}(1/\gamma, \gamma)$. Note that by our choice of γ' via Claim 4.6, and because $\gamma \leq \gamma'$, there are no i, j, k such that $(V_i, V_j), (V_j, V_k), (V_i, V_k)$ are γ -regular and $d(V_i, V_j), d(V_j, V_k), d(V_i, V_k) \geq \delta$ because such sets span at least one triangle (in fact, many). As by item (iii) the densities of the instances in \mathcal{R} are taken from $\{0, \gamma, 2\gamma, \dots, 1\}$ we can make at most $\gamma n^2 \leq \delta n^2$ changes and “round down” the densities between the sets into a multiple of γ , while maintaining the regularity of the regular-pairs (we can use Lemma 3.1 here). This means that the new graph satisfies a regularity-instance $R \in \mathcal{R}$, which means that G was δ -close to satisfying R . \square

Our second application of Theorem 2 is concerned with testing k -colorability. This property was first implicitly proved to be testable in [32]. Much better upper bounds were obtained in [24], and further improved by [5]. As in the case of H -freeness, the main ideas of the proofs in [32, 24, 5] is that if G is ϵ -far from being k -colorable then a large enough sample of vertices will not be k -colorable with high probability. Here we derive this result by applying Theorem 2. Though we derive here only the testability of k -colorability, simple variants of the argument can be used to show that all the

partition-problems studied in [24] are testable¹.

COROLLARY 7.2. *k -colorability is testable.*

PROOF. (SKETCH): By Theorem 2 it is enough to show that k -colorability is regular-reducible. Fix any $\delta > 0$ and define \mathcal{R} to be all the regularity-instances R satisfying the following: (i) They have regularity measure δ (ii) They have order at least $1/\delta$ and at most $T_{2.3}(2/\delta, \delta)$ (iii) Their densities $\eta_{i,j}$ are taken from $\{0, \delta, 2\delta, \dots, 1\}$. (iv) The following graph $T = T(R)$ is k -colorable: if R has order t then T has t vertices, and $(i, j) \in E(T)$ iff $\eta_{i,j} > 0$.

To show that this is a valid reduction, assume first that G is ϵ -far from being k -colorable. Assume G is $(\epsilon - \delta)$ -close to satisfying a regularity instance $R \in \mathcal{R}$. We can thus make $(\epsilon - \delta)n^2$ edge modifications and get a graph satisfying R . We also remove all edges inside the sets V_i . As by item (ii) each set has size at most δn we remove less than δn^2 edges. The total number of edges removed is thus less than ϵn^2 . By property (iv) of the regularity instances of \mathcal{R} this means that the new graph is k -colorable, which is impossible because we made less than ϵn^2 edge modifications and G was assumed to be ϵ -far from being k -colorable. Assume now that G is k -colorable and let V_1, \dots, V_k be the partition of $V(G)$, which is determined by a legal k -coloring of G . Break every set V_i into sets $U_{i,1}, \dots, U_{i,2/\delta k}$ of size $\frac{1}{2}\delta n$. Put all the leftovers from each set in another set L of size $\frac{1}{2}\delta n$. By Lemma 2.3, starting from this equipartition we can get a δ -regular equipartition of G of order at most $T_{2.3}(2/\delta, \delta)$. Note that disregarding the refinement of L the new equipartition must satisfy item (v) in the definition of \mathcal{R} . As by item (iii) the densities of the instances in \mathcal{R} are taken from $\{0, \delta, 2\delta, \dots, 1\}$ we can make at most δn^2 edge modifications and thus “round down” the densities between the sets into a multiple of δ , while maintaining the regularity of the regular-pairs (we can use Lemma 3.1 here). This means that the new graph satisfies a regularity-instance $R \in \mathcal{R}$, which means that G was δ -close to satisfying R . \square

The examples that were discussed above apply Theorem 2 to obtain positive results. Our third application of Theorem 2 derives a negative result. The main focus of [15] is testing for isomorphism to a given fixed graph. It shows that the query complexity of testing for isomorphism grows with a certain parameter, which measures the “complexity” of the graph. Without going into details we just mention that under this measure random graphs are complex. Here we prove that testing for being isomorphic to a graph generated by $G(n, 0.5)$ requires a super-constant number of queries.

COROLLARY 7.3. *Let I be a graph generated by $G(n, 0.5)$. Then, with probability $1 - o(1)$ the property of being isomorphic to I is not testable.*

PROOF. (SKETCH): By Theorem 2 it is enough to show that with probability $1 - o(1)$ the property of being isomorphic to I is not regular-reducible. Note, that now there is only one value of n to consider in Definition 2.5 because the property we consider is a property of n -vertex graphs. Consider a graph generated by $G(n, 0.5)$. Clearly, by a simple Chernoff bound the bipartite graph on any pair of sets

¹An alert reader may note that our proof of Theorem 2 applies the result of [17], which relies on the result of [24]. Thus, in the strict sense it is wrong to say that we infer the result of [24] from ours. However, it is not difficult to see that the result of [24] also follows from our (self-contained) proof of Lemma 5.2.

of vertices of size \sqrt{n} has density ≈ 0.5 . We claim that if I satisfies this property then it is not regular-reducible. Suppose it is regular-reducible and consider a small δ , say $\delta = 0.01$. Let \mathcal{R} be the set of regularity-instances, which corresponds to this value of δ . Let G be a graph isomorphic to I . By Definition 2.5 it must be the case that G is δ -close to satisfying some $R \in \mathcal{R}$. By the properties of I this means that most densities of R must be close to 0.5. Let k denote the order of R and let $\eta_{i,j}$ denote its densities. Consider a random k -partite graph on sets of vertices V_1, \dots, V_k each of size n/k , where the bipartite graph connecting V_i and V_j is a random bipartite graph with edge density $\eta_{i,j}$. Clearly this graph is δ -close to satisfying R . On the other hand, it is not difficult to see that as most of the densities $\eta_{i,j}$ should be close to 0.5, then with high probability such a graph must be α -far from being isomorphic to I , for some fixed $\alpha > 0$, say $\alpha = 0.03$. This means that we have a graph that is 0.03-far from satisfying the property and is yet 0.01-close to satisfying one of the regularity-instances of \mathcal{R} . As $\delta = 0.01$, this violates the second condition of Definition 2.5. \square

8. CONCLUDING REMARKS

The main result of this paper gives a combinatorial characterization of the graph properties, which can be tested with a constant number of edge queries in the dense graph model, possibly with a two-sided error. Together with the (near) characterization of [7] of the graph properties that can be tested with one-sided error, and the result of [17] showing that any testable property is also estimable, we get a more or less complete answer to many of the *qualitative* questions on testing graph properties in the dense model. While property testing in the dense model is relatively well understood, there are no general positive or negative results on testing graph properties in the bounded-degree model [26] or the general density model [30]. In these models the query complexity of the tester usually depends on the size of the input. It seems interesting and challenging to obtain general results in these models. One interesting problem is which of the partition problems which were studied in [24] can be tested using a sublinear number of queries. It will also be very interesting to give general positive and negative results concerning the testing of boolean functions.

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