# [Re] Diffusion-Based Adversarial Sample Generation for Improved Stealthiness and Controllability

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# **Reproducibility Summary**

- 2 Template and style guide to ML Reproducibility Challenge 2020. The following section of Repro-
- 3 ducibility Summary is mandatory. This summary must fit in the first page, no exception will be
- 4 allowed. When submitting your report in OpenReview, copy the entire summary and paste it in the
- 5 abstract input field, where the sections must be separated with a blank line.

#### 6 Scope of Reproducibility

- 7 State the main claim(s) of the original paper you are trying to reproduce (typically the main claim(s)
- 8 of the paper). This is meant to place the work in context, and to tell a reader the objective of the
- 9 reproduction.

#### 10 Methodology

- Briefly describe what you did and which resources you used. For example, did you use author's code?
- 12 Did you re-implement parts of the pipeline? You can also use this space to list the hardware used,
- and the total budget (e.g. GPU hours) for the experiments.

#### 14 Results

- 15 Start with your overall conclusion where did your results reproduce the original paper, and where
- did your results differ? Be specific and use precise language, e.g. "we reproduced the accuracy to
- within 1% of reported value, which supports the paper's conclusion that it outperforms the baselines".
- 18 Getting exactly the same number is in most cases infeasible, so you'll need to use your judgement to
- decide if your results support the original claim of the paper.

#### 20 What was easy

- 21 Describe which parts of your reproduction study were easy. For example, was it easy to run the
- 22 author's code, or easy to re-implement their method based on the description in the paper? The goal
- 23 of this section is to summarize to a reader which parts of the original paper they could easily apply to
- 24 their problem.

# What was difficult

- Describe which parts of your reproduction study were difficult or took much more time than you
- 27 expected. Perhaps the data was not available and you couldn't verify some experiments, or the
- 28 author's code was broken and had to be debugged first. Or, perhaps some experiments just take too
- much time/resources to run and you couldn't verify them. The purpose of this section is to indicate
- to the reader which parts of the original paper are either difficult to re-use, or require a significant
- amount of work and resources to verify.

32	Communication	with	original	authors

Briefly describe how much contact you had with the original authors (if any).

- 34 The following section formatting is optional, you can also define sections as you deem fit.
- 35 Focus on what future researchers or practitioners would find useful for reproducing or building
- 36 upon the paper you choose.

## 1 Introduction

- 38 A few sentences placing the work in high-level context. Limit it to a few paragraphs at most; your
- report is on reproducing a piece of work, you don't have to motivate that work.

# 40 2 Scope of reproducibility

- 41 Introduce the specific setting or problem addressed in this work, and list the main claims from the
- original paper. Think of this as writing out the main contributions of the original paper. Each claim
- should be relatively concise; some papers may not clearly list their claims, and one must formulate
- them in terms of the presented experiments. (For those familiar, these claims are roughly the scientific
- 45 hypotheses evaluated in the original work.)
- 46 A claim should be something that can be supported or rejected by your data. An example is,
- 47 "Finetuning pretrained BERT on dataset X will have higher accuracy than an LSTM trained with
- 48 GloVe embeddings." This is concise, and is something that can be supported by experiments. An
- 49 example of a claim that is too vague, which can't be supported by experiments, is "Contextual
- 50 embedding models have shown strong performance on a number of tasks. We will run experiments
- evaluating two types of contextual embedding models on datasets X, Y, and Z."
- This section roughly tells a reader what to expect in the rest of the report. Clearly itemize the claims you are testing:
  - Diffusion-Based Projected Gradient Descent (Diff-PGD) generates realistic adversarial samples
    - Diff-PGD can be applied to specific tasks such as digital attacks, physical-world attacks, and style-based attacks, outperforming baseline methods such as PGD, AdvPatch, and AdvCam.
    - Diff-PGD is more stable and controllable compared to existing methods for generating natural-style adversarial samples.
    - Diff-PGD surpasses the original PGD in Transferability and Purification power
- Each experiment in Section 4 will support (at least) one of these claims, so a reader of your report should be able to separately understand the *claims* and the *evidence* that supports them.

## 63 Methodology

- Explain your approach did you use the author's code, or did you aim to re-implement the approach
- from the description in the paper? Summarize the resources (code, documentation, GPUs) that you
- 66 used.

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#### 67 3.1 Model descriptions

Include a description of each model or algorithm used. Be sure to list the type of model, the number of parameters, and other relevant info (e.g. if it's pretrained).

## o 3.2 Datasets

- For each dataset include 1) relevant statistics such as the number of examples and label distributions,
- 2) details of train / dev / test splits, 3) an explanation of any preprocessing done, and 4) a link to
- 73 download the data (if available).

#### 74 3.3 Hyperparameters

- 75 Describe how the hyperparameter values were set. If there was a hyperparameter search done, be
- sure to include the range of hyperparameters searched over, the method used to search (e.g. manual
- 577 search, random search, Bayesian optimization, etc.), and the best hyperparameters found. Include the
- 78 number of total experiments (e.g. hyperparameter trials). You can also include all results from that
- 79 search (not just the best-found results).

#### 80 3.4 Experimental setup and code

- 81 Include a description of how the experiments were set up that's clear enough a reader could replicate
- 82 the setup. Include a description of the specific measure used to evaluate the experiments (e.g. accuracy,
- precision@K, BLEU score, etc.). Provide a link to your code.

#### 84 3.5 Computational requirements

- 85 Include a description of the hardware used, such as the GPU or CPU the experiments were run on.
- For each model, include a measure of the average runtime (e.g. average time to predict labels for a
- 87 given validation set with a particular batch size). For each experiment, include the total computational
- 88 requirements (e.g. the total GPU hours spent). (Note: you'll likely have to record this as you run
- 89 your experiments, so it's better to think about it ahead of time). Generally, consider the perspective of
- a reader who wants to use the approach described in the paper list what they would find useful.

#### 91 4 Results

- 92 Start with a high-level overview of your results. Do your results support the main claims of the
- 93 original paper? Keep this section as factual and precise as possible, reserve your judgement and
- 94 discussion points for the next "Discussion" section.

# 95 4.1 Results reproducing original paper

- 96 For each experiment, say 1) which claim in Section 2 it supports, and 2) if it successfully reproduced
- 97 the associated experiment in the original paper. For example, an experiment training and evaluating a
- model on a dataset may support a claim that that model outperforms some baseline. Logically group
- 99 related results into sections.

#### 100 4.1.1 Result 1

#### 101 **4.1.2 Result 2**

## 102 4.2 Results beyond original paper

- Often papers don't include enough information to fully specify their experiments, so some additional
- experimentation may be necessary. For example, it might be the case that batch size was not specified,
- and so different batch sizes need to be evaluated to reproduce the original results. Include the results
- of any additional experiments here. Note: this won't be necessary for all reproductions.

# 107 4.2.1 Additional Result 1

#### 108 4.2.2 Additional Result 2

#### 109 5 Discussion

- 110 Give your judgement on if your experimental results support the claims of the paper. Discuss the
- strengths and weaknesses of your approach perhaps you didn't have time to run all the experiments,
- or perhaps you did additional experiments that further strengthened the claims in the paper.

#### 113 5.1 What was easy

- Give your judgement of what was easy to reproduce. Perhaps the author's code is clearly written and
- easy to run, so it was easy to verify the majority of original claims. Or, the explanation in the paper
- was really easy to follow and put into code.
- 117 Be careful not to give sweeping generalizations. Something that is easy for you might be difficult
- to others. Put what was easy in context and explain why it was easy (e.g. code had extensive API
- documentation and a lot of examples that matched experiments in papers).

#### 120 5.2 What was difficult

- List part of the reproduction study that took more time than you anticipated or you felt were difficult.
- Be careful to put your discussion in context. For example, don't say "the maths was difficult to
- follow", say "the math requires advanced knowledge of calculus to follow".

## 5.3 Communication with original authors

- Document the extent of (or lack of) communication with the original authors. To make sure the
- reproducibility report is a fair assessment of the original research we recommend getting in touch
- 127 with the original authors. You can ask authors specific questions, or if you don't have any questions
- you can send them the full report to get their feedback before it gets published.

#### 129 References

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