### **Dream Fusion**: Text-to-3D using 2D Diffusion

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## **Quick Summary**

- using a pretrained 2D text-to-image diffusion model (Google's Imagen) to perform 3D synthesis
- no need in 3D training data
- no need to make modifications in Imagen



# Before & after projects

	text-to-image	text-to-3D	text-to-video
January 2021	DALL-E (Open AI)		
April 2022	DALL-E 2 (Open Al)		
May 2022	lmagen (Google)		
August 2022	Stable Diffusion (Stability Al)		
September 2022		Dream Fusion / Get3D (Google) (Nvidia)	Make-A-Video (Meta Al)
November 2022	Stable Diffusion Version 2 (Stability Al)		
December 2022		3D Avatar Diffusion (Microsoft)	

#### Competitors

Dream Fusion (Google)



3D training data: none

quality: bad

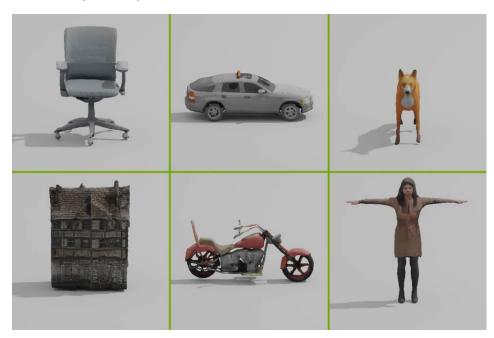
areas of application: few (for now) / many (in future)

• construction composition of objects: **good** 

• open source: none

• free-to-use: none

#### GET3D (Nvidia)



• 3D training data: **ShapeNet dataset** 

quality: good

areas of application: many

construction composition of objects: **none** 

• open source: **yes** 

free-to-use: **yes** 

#### Competitors

Dream Fusion (Google)



3D training data: none

quality: bad

areas of application: few (for now) / many (in future)

construction composition of objects: good

• open source: **none** 

free-to-use: none

3D Avatar Diffusion (Microsoft)



• 3D training data: Face Synthetics dataset

quality: good

• areas of application: **few** (only human avatars)

construction composition of objects: good

• open source: none

free-to-use: none

### **Strength** of Dream Fusion

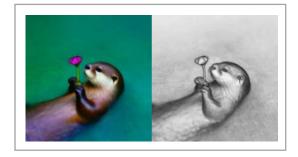
- 1. generated 3D objects correspond to the text input well
- 2. capable of creating not just a singular 3D objects but entire 3D scenes
- 3. realistic interactions and intersections of objects



difficult materials



objects composition / intersection



volumetric effects based on environment

#### Weaknesses of Dream Fusion

1. **quality issues**: blurry, dark, unnatural high-contrast colors



2. **multi-facing problem** (a.k.a. Janus problem)

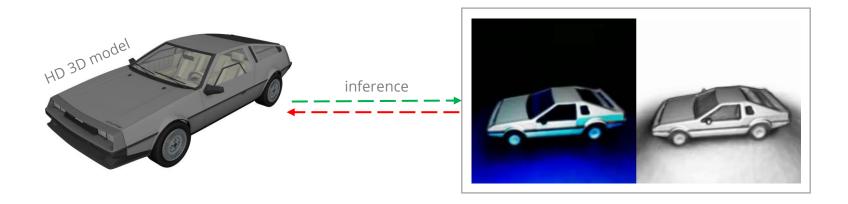




- a DSLR photo of a squirrel + wearing a purple hoodie
- a DSLR photo of a squirrel + wearing a purple hoodie + eating a hamburger

#### Additional experiments ideas

- 1. need a more lightweight version of **Dream Fusion** that can ran quickly on your local machine with low RAM
- 2. generate multiple outputs by a single input: how can we generate multiple 3D scenes with the same text?
- 3. adding into the training pipeline ground truth objects in the form of manually created high quality 3D models



## Application Idea (right now)

Generated by **Dream Fusion** 3D models can be used in Al training applications / simulators

- **fact:** for training self-driving cars developers use 3D simulators, which should be filled with different types of 3D assets: *trees, roads, pedestrians, other cars, motorcycles, bikes, houses and etc.*
- task: object recognition by images + lidar scans
- **problem:** simulators have limited and identical packs of 3D assets
- consequence: overfitting of Al system (bad accuracy)
- **solution:** generate many different 3D assets to increase diversity
- result: removed overfitting



#### Sources

- 1. [2209.14988] DreamFusion: Text-to-3D using 2D Diffusion (arxiv.org)
- 2. <u>DreamFusion: Text-to-3D using 2D Diffusion (dreamfusion3d.github.io)</u>
- 3. <u>DreamFusion: Text-to-3D using 2D Diffusion! YouTube</u>
- 4. Google's DreamFusion AI: Text to 3D YouTube
- 5. <u>DreamFusion: Text-to-3D using 2D Diffusion | OpenReview</u>