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Label Freedom: Stable diffusion on Remote Sensing Semantic **Segmentation Data Generation**

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Background

Since 2023, AIGC (artificial intelligence generated content) has shown the performance of AGI (artificial general intelligence) seed. Furthermore, SOTA of supervised downstream task is close to saturation. We took the advantage of stable diffusion, proposed a data generation pipeline. This approach generated semantic segmentation label and image data, and the generated data achieved high accuracy as 0.516 mIoU in LoveDA dataset, which closed to the original dataset result as 0.530 mIoU.

- 1. We implemented the stable diffusion algorithm in remote sensing satellite image generation.
- 2. We proposed a semantic segmentation image-label pair generation approach based on LoRA and stable diffusion.
- 3. The generated datasets were validated by Mask2Former algorithm, which achieved high accuracy as 0.516 mIoU in LoveDA dataset, closed to the original dataset.



LoveDA dataset:

2522 train, 1669 val, 1796 test, 5987 total. 1024 x 1024, 0.3m resolution

B.g., building, road, water, barren, forest, agriculture.

Workflow





LoRA + Stable diffusion Label Generation



Prompt: "professional photograph, Satellite imagery, remote sensing, worldview"

ControlNet + Stable diffusion Image generation





a). Stable diffusion based remote sensing



Mask2Former, Generation quality validation

Data augmentation in training and testing

semantic segmentation dataset generation, start from noise, restrain by LoRA and ControlNet; **b**). Quantitative generated dataset quality validation based on Mask2Former semantic segmentation algorithm.

Result









Real image



As shown on the left images, this approach could be used to generated aerial image in different style and resolution, the scale parameter could be used to control the image fineness.

The generated images have similar performance to manually annotation, thus time and manpower could be saved by our approach. This approach could be generalized to instance segmentation, and for the street view image generation, 3D modeling, etc.

Semantic segmentation performance comparison on generated data and original data

Data	background	building	road	water	barren	forest	agriculture	mIoU	
Gen.	0.435	0.594	0.520	0.811	0.159	0.466	0.627	0.516	
Ori.	0.464	0.591	0.595	0.818	0.153	0.472	0.615	0.530	
Gen. + Ori.	0.466	0.579	0.601	0.812	0.179	0.467	0.629	0.533	











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