Supplementary Materials for CookGAN: Causality based Text-to-Image Synthesis

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1. Additional Generation Examples

1.1. Content Manipulability

Recipe Manipulation	Concrete Operation	GT	Before Manipulation	After Manipulation
Add Ingredients	+ Carrot			Contraction of the second
		X		
Minus Ingredients	- Carrot			
			A	
Replace Ingredients	Carrot → lettuce			

Figure 1. Examples of generated images by CookGAN with different recipe manipulation operators.

1.2. Visual Quality



Figure 2. Comparison of food images generated by CookGAN, StepGAN, IngredientGAN and StackGAN++.

2. Additional IS Results

We also compute Inception score (IS) in different image resolution. At the resolution of 128×128 , the IS of CookGAN is 6.1% better than ACME [1]. At the resolution of 64×64 , CookGAN is better than R^2GAN [2] by 7.2%.

References

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- [2] Bin Zhu, Chong-Wah Ngo, Jingjing Chen, and Yanbin Hao. R2GAN: Cross-modal recipe retrieval with generative adversarial network. In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*, pages 11477–11486, 2019.