

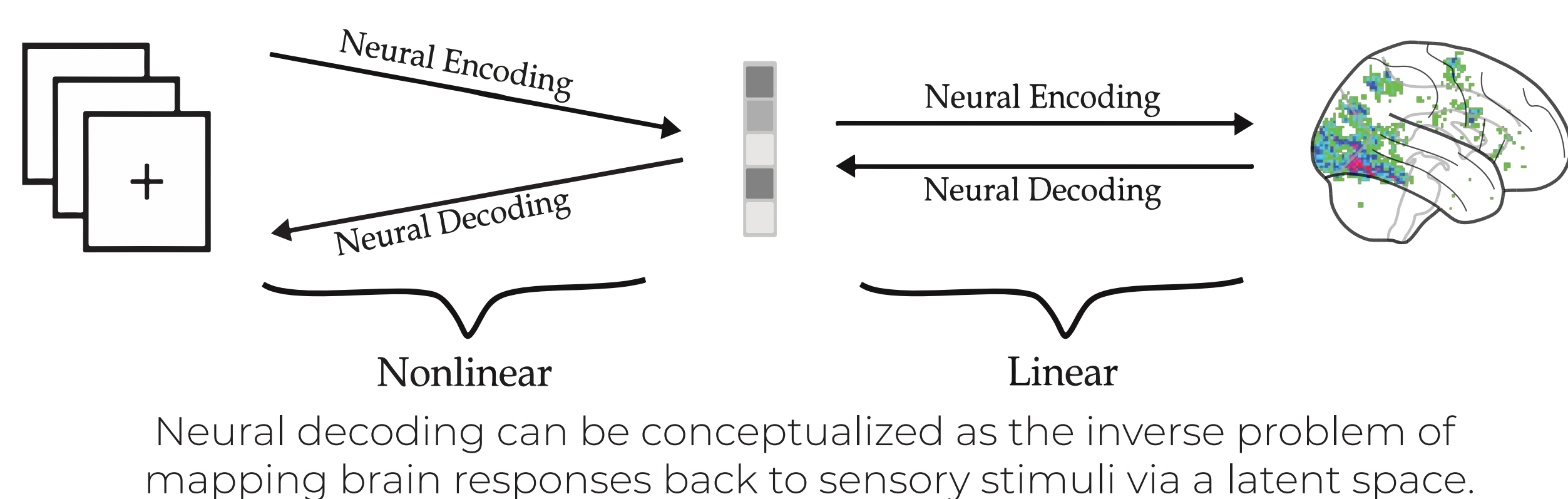
Hyperrealistic neural decoding: Reconstruction of face stimuli from fMRI measurements via the QAN latent space

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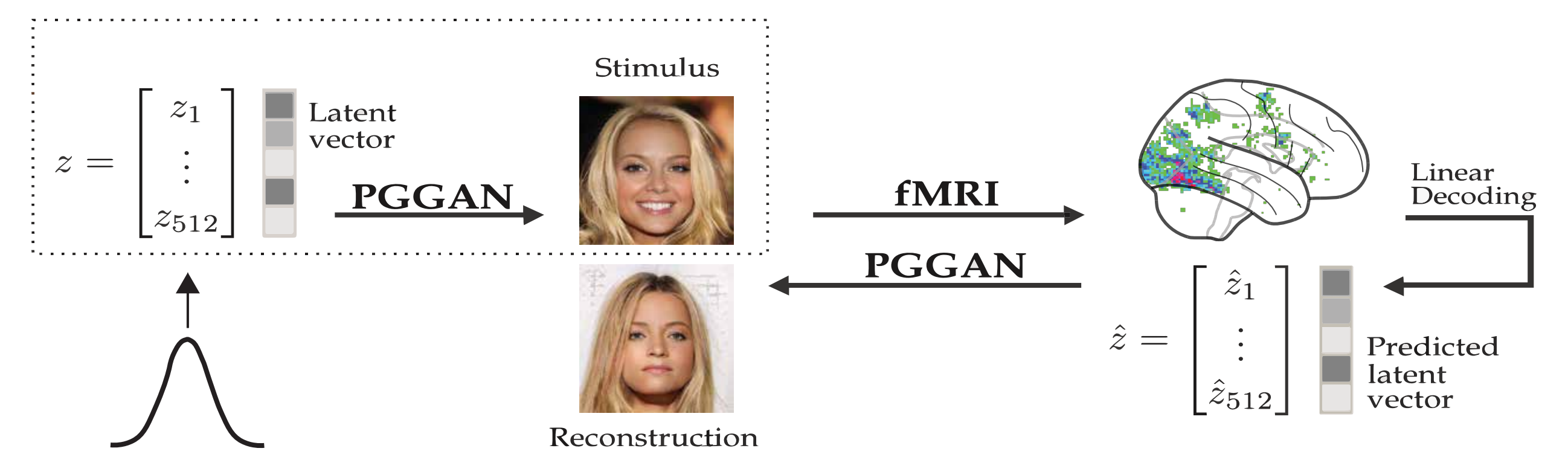
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INTRODUCTION

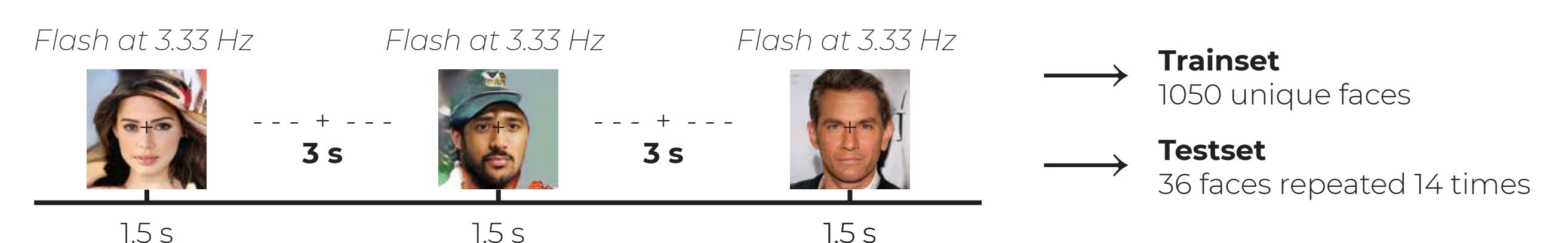
- Unlike their supervised counterparts [1], more biologically plausible unsupervised deep neural networks (DNNs) have paradoxically been unsuccessful in modeling neural representations [2].
- At the same time, generative adversarial networks (GANs) have become one of the most powerful unsupervised DNNs in modeling image representations.
- Problem:** GANs have high potential in modeling neural representations, but testing this hypothesis is not possible because latent representations cannot be obtained retrospectively [3].
- Solution:** A novel experimental paradigm for well-controlled yet naturalistic stimuli with known latent representations, and a GAN-based neural decoding model for HYperrealistic reconstruction of PERception (HYPER) with unprecedented accuracy to date.



METHODS



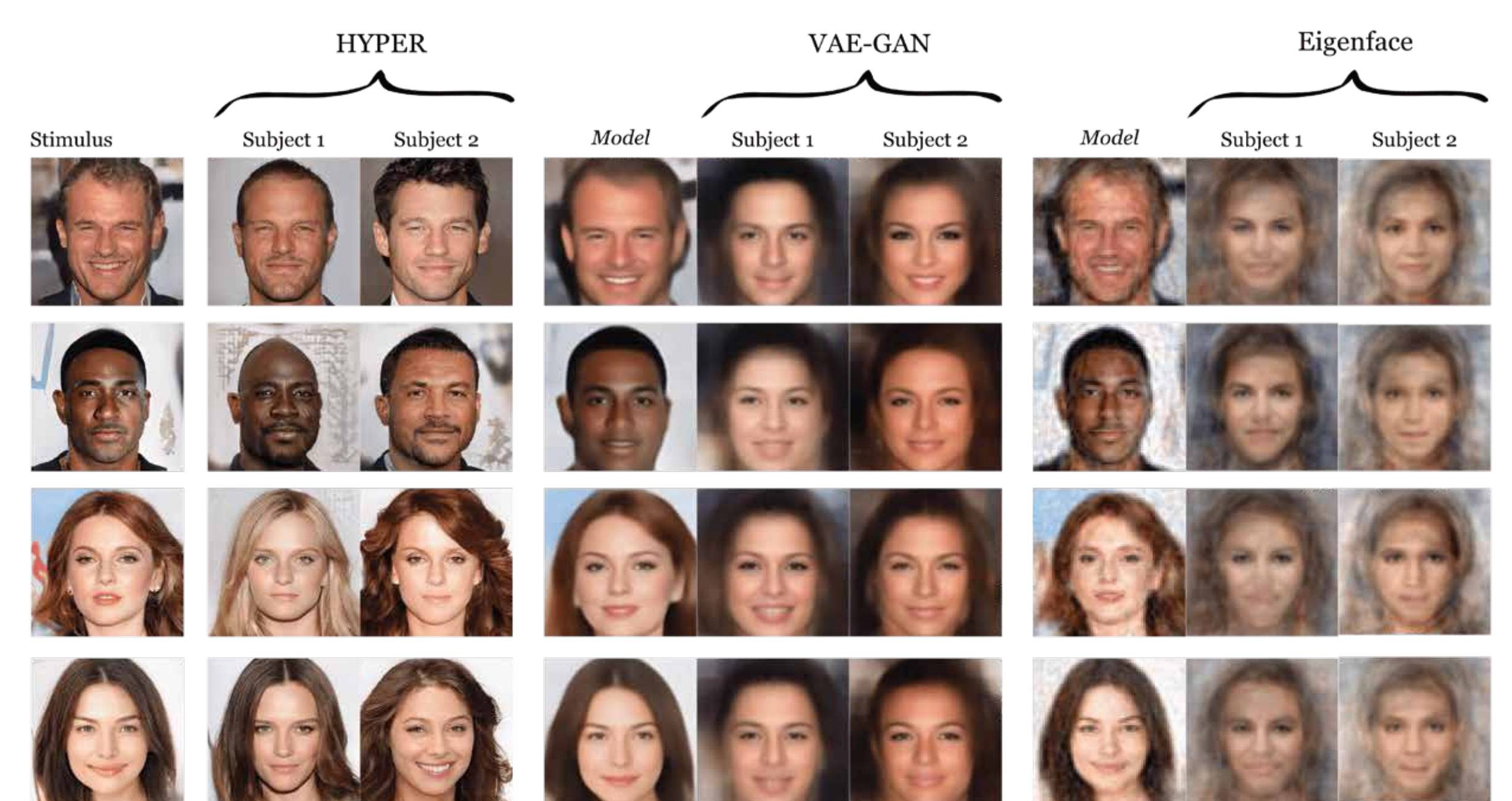
- The progressive growing of GANs (PGGAN) model [4] generates photo-realistic faces (1024 x 1024 pixels) that resemble celebrities from randomly sampled latent vectors (512 dims).
- Blood-oxygen-level dependent hemodynamic responses (TR = 1.5 s, voxel size = 2 x 2 x 2 mm³, whole brain) of two subjects were measured during presentation of faces.
- Linear decoding by a dense layer at the beginning of PGGAN to transform brain data into latent vectors.



RESULTS

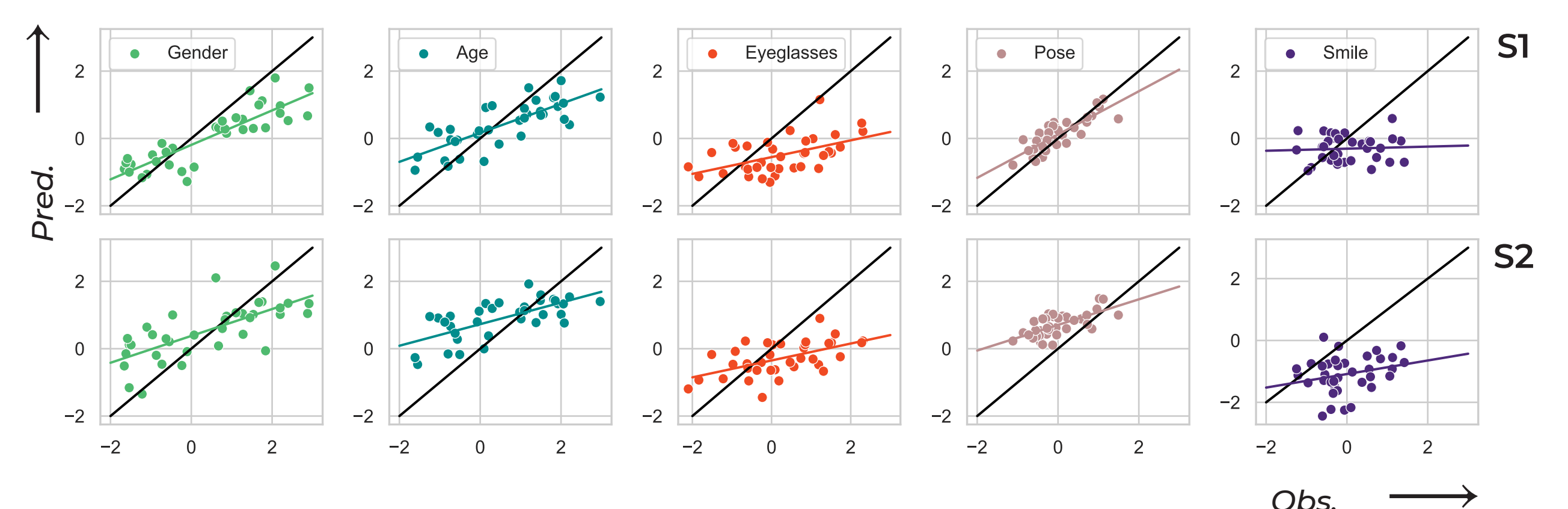


QUALITATIVE RESULTS



QUANTITATIVE RESULTS

		Lat. sim.	Feat. sim.	Struct. sim.
S1	HYPER	0.4521 ± 0.0026	0.1745 ± 0.0038	0.6663 ± 0.0115
	VAE-GAN	(p < 0.001; perm.test)	(p < 0.001; perm.test)	(p < 0.001; perm.test)
	Eigenface	-	0.1416 ± 0.0025	0.5598 ± 0.0151
S2	HYPER	0.4447 ± 0.0020	0.1715 ± 0.0049	0.6035 ± 0.0128
	VAE-GAN	(p < 0.001; perm.test)	(p < 0.001; perm.test)	(p < 0.001; perm.test)
	Eigenface	-	0.1461 ± 0.0022	0.5832 ± 0.0141



CONCLUSIONS

With the introduced paradigm and model we:

- Showed that unsupervised deep neural networks can successfully model neural representations of naturalistic stimuli
- Showed that the GAN latent space approximates the neural face manifold
- Obtained state-of-the-art reconstructions of perceived faces from brain activations

Considering the speed of progress in the field of generative modeling, the HYPER framework will likely result in even more impressive reconstructions of perception in the near future.

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