

# Computational Statistics

## *Hyperspherical VAE*

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2. Sampling method
3. Reparameterization Trick
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5. Conclusion and Discussion



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introduire la méthode et les contributions [DFDC<sup>+</sup>]



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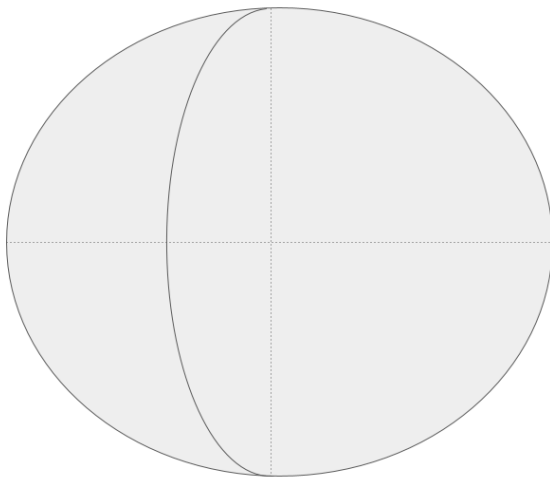
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Démontrer que la méthode de sampling marche



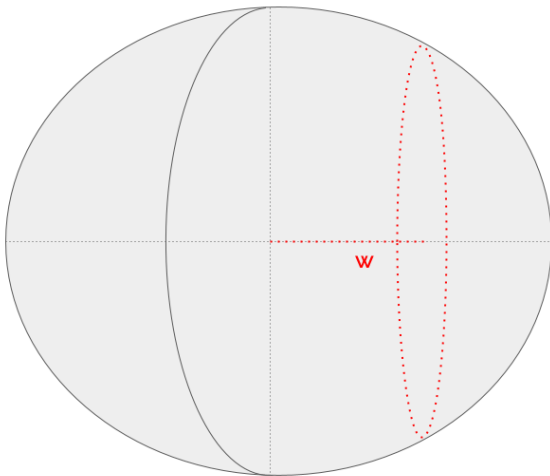
# Sampling $w$ from $g(w|\kappa, \theta)$



$S^2$  : unit sphere in  $\mathbb{R}^3$

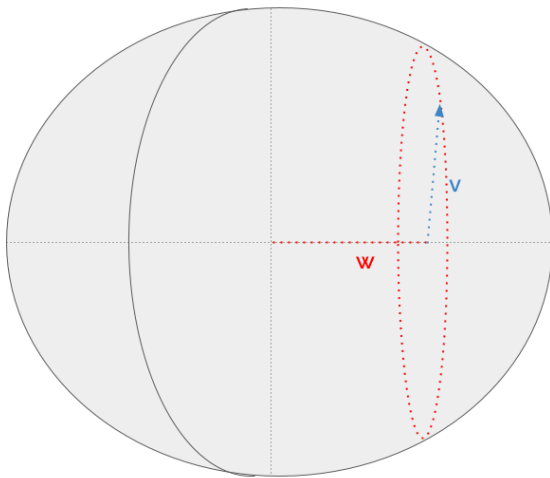


# Sampling $w$

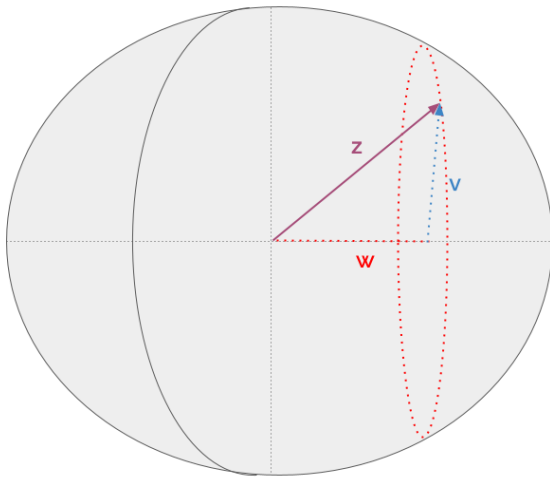




# Sampling $w$



# Sampling $w$



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# Reparameterization Trick

The authors use a reparameterization trick that has been extended to distributions that can be sampled using rejection sampling [NRLB20].

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**Algorithm 1** Reparameterized Rejection Sampling (from [NRLB20])

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```
1:  $i \leftarrow 0$ 
2: repeat
3:    $i \leftarrow i + 1$ 
4:   Propose  $\varepsilon_i \sim s(\varepsilon)$ 
5:   Simulate  $u_i \sim \mathcal{U}[0, 1]$ 
6: until  $u_i < \frac{g(h(\varepsilon_i, \theta); \theta)}{r(h(\varepsilon_i, \theta); \theta)}$ 
7: return  $\varepsilon_i$ 
```

---



By noting  $\pi(\varepsilon|\theta)$  the distribution of the resulting  $\varepsilon$ , we have

$$\nabla_{\theta} \mathbb{E}_{g(\varepsilon|\theta)}[\dots] = \mathbb{E}_{\pi(\varepsilon|\theta)}[\dots] = \mathbb{E}_{(\varepsilon_i, U_i)_i}[\dots]$$

Problem:  $(\varepsilon_i, U_i)_{i \in \mathbb{N}}$  is not a random variable (it is a stochastic process)  
No reference to a convergence proof in [DFDC<sup>+</sup>, NRLB20, PBJ12, MG14]



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# Experiments on link prediction

reproduire l'expérience

- data (Ines)
- implementer les modèles (Victor VGAE)
- entraînement et evaluation



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Limitations à la fin : discussion sur la grande dimension  
*vanishing surface problem*  
vérifier différentes dimension de l'espace latent  
et algo vraiment utile en petite ou moyenne dimension ?



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Hyperspherical variational auto-encoders.
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Neural variational inference and learning in belief networks, 2014.
-  Christian A. Naesseth, Francisco J. R. Ruiz, Scott W. Linderman, and David M. Blei.  
Reparameterization gradients through acceptance-rejection sampling algorithms, 2020.
-  John Paisley, David Blei, and Michael Jordan.  
Variational bayesian inference with stochastic search, 2012.

