

# On the effectiveness of weight-encoded neural implicit 3D shapes

LIRIS - April/August 2022 - Nissim Maruani

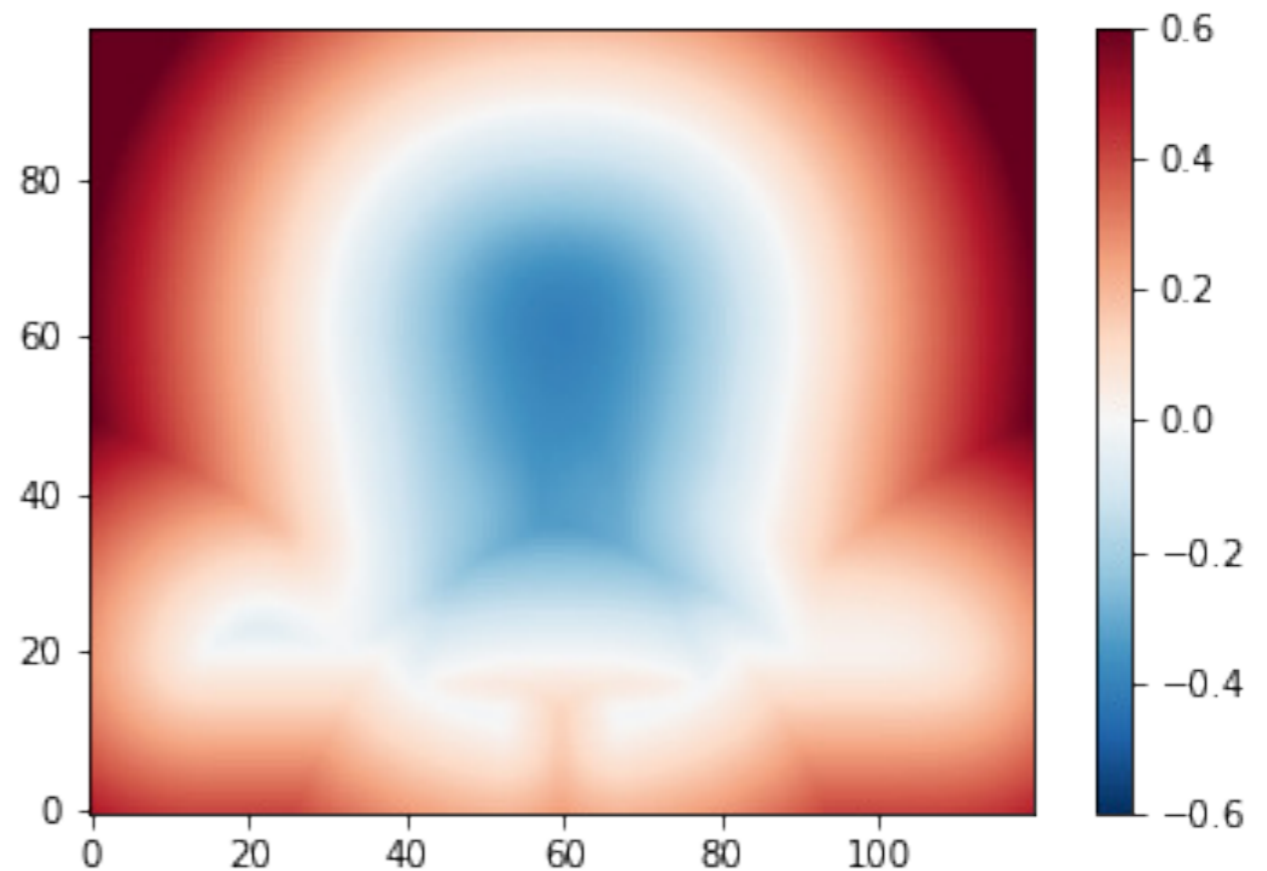
Davies, T., Nowrouzezahrai, D., & Jacobson, A. (2020). *On the effectiveness of weight-encoded neural implicit 3D shapes*. arXiv preprint arXiv:2009.09808.

# Background: SDF



**3D model**

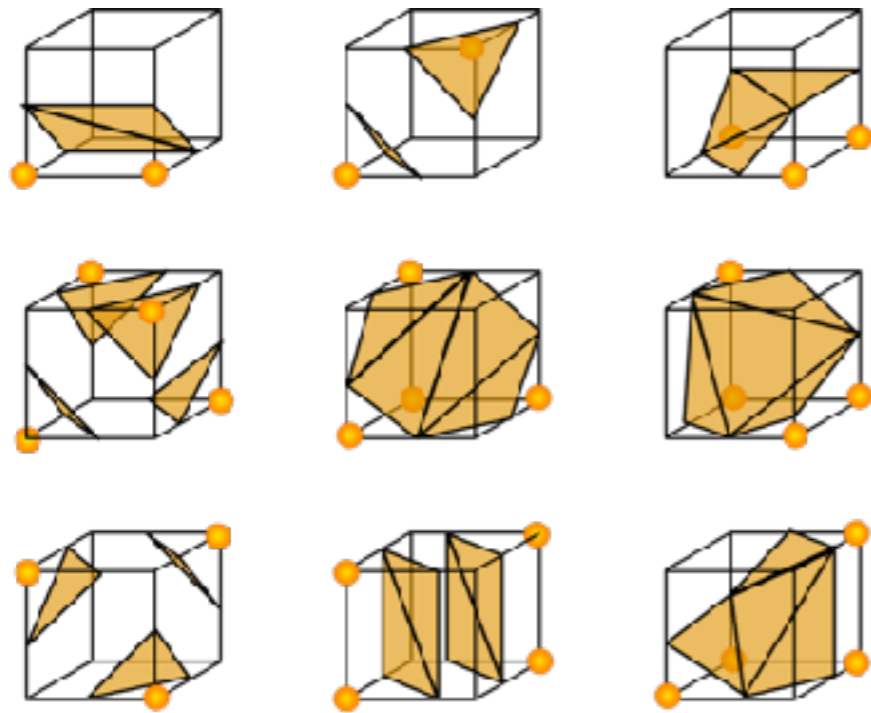
Closed volume  $\Omega$



**Signed Distance Function**

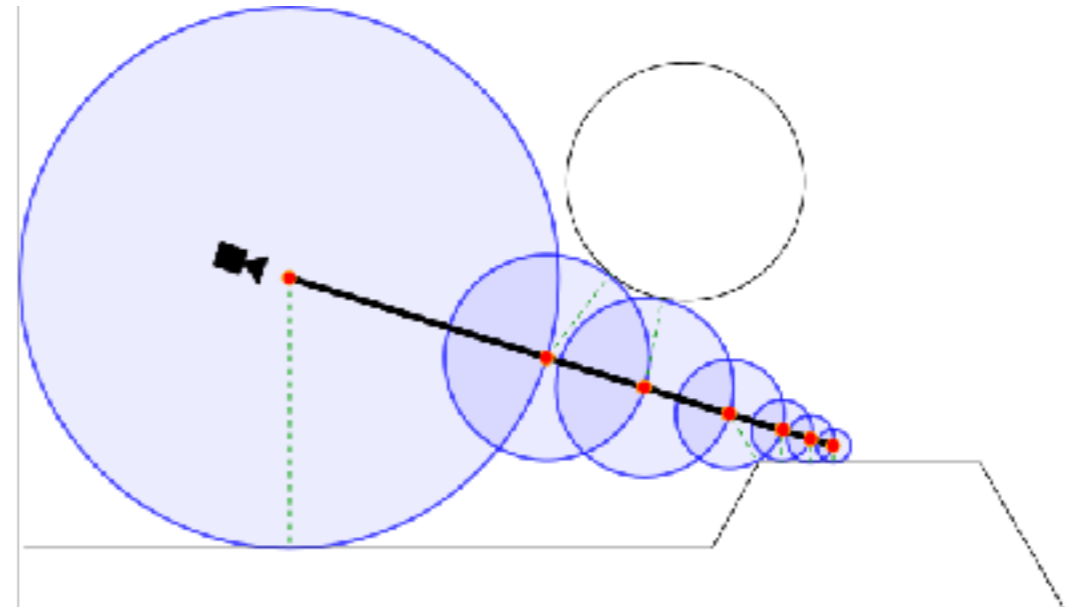
$$f : x \in \mathbb{R}^3 \mapsto (-1)^{1_{x \in \Omega}} d(x, \partial\Omega)$$

# Background: SDF



**Marching cubes**

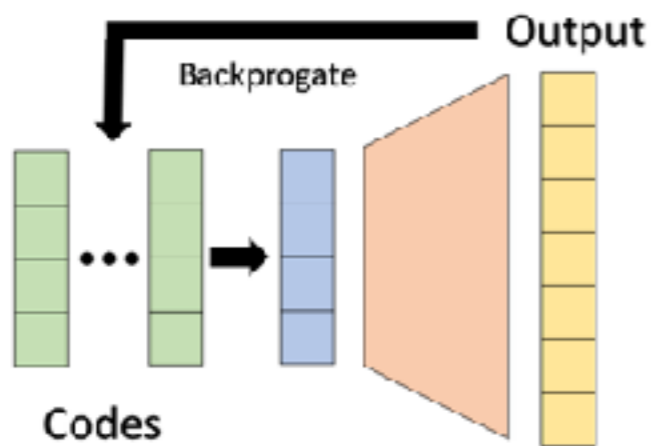
Surface extraction



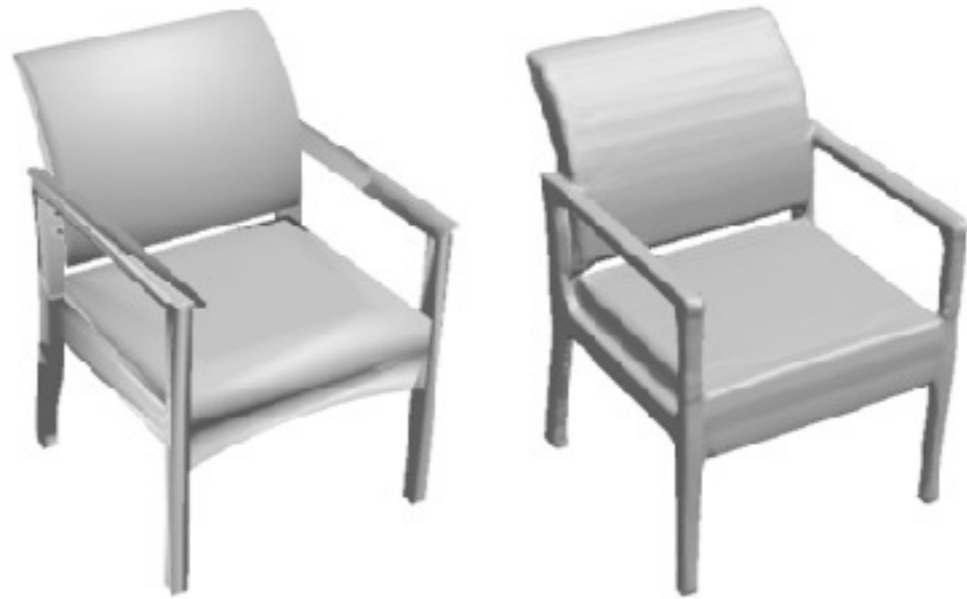
**Sphere tracing**

Rendering

# Background: DeepSDF



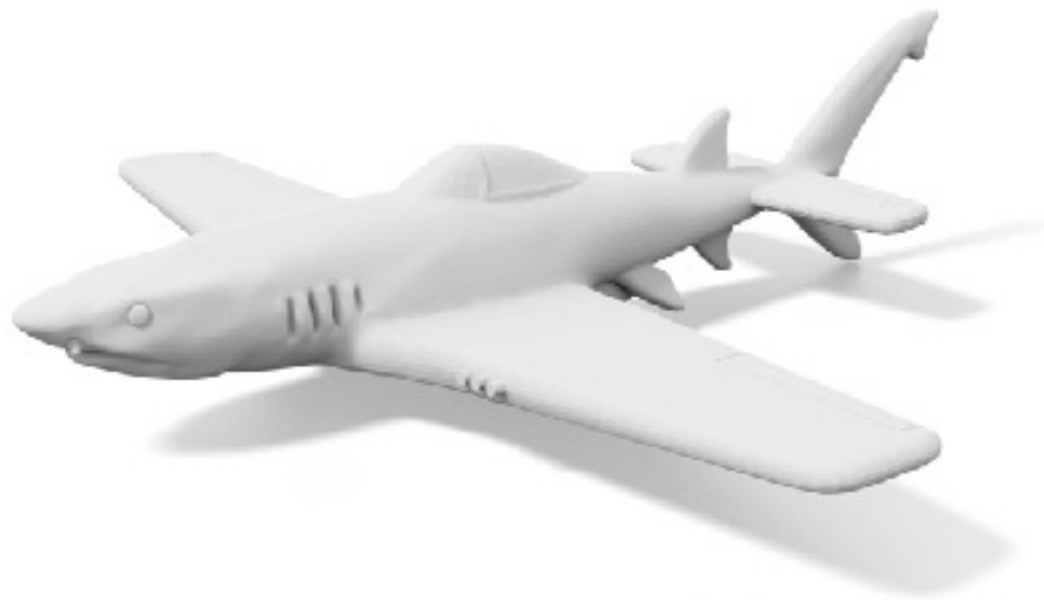
Network  $f(x_i, z_i) = s_i$



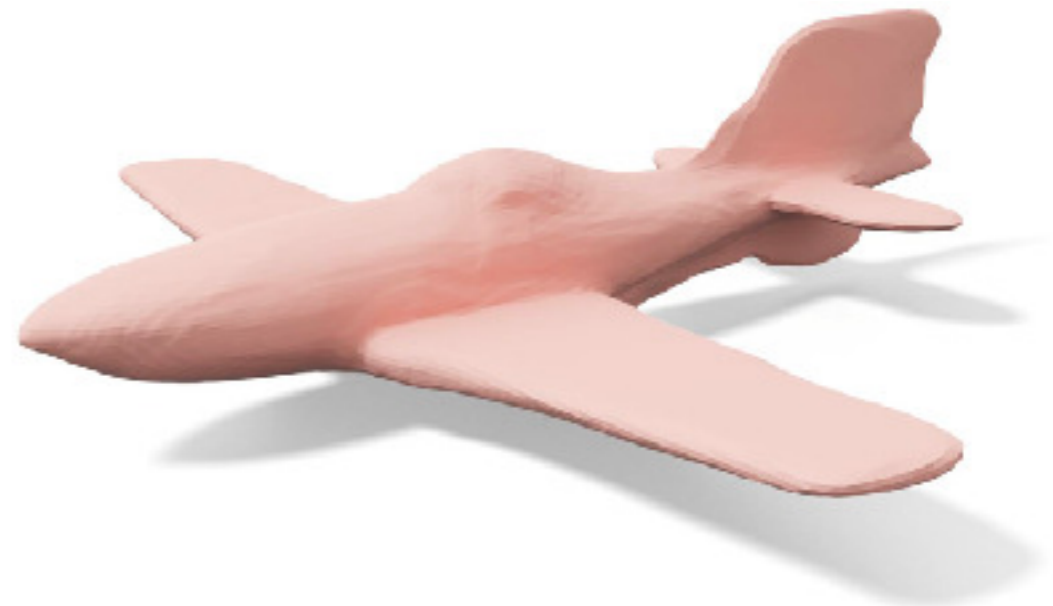
Ground Truth    Reconstruction

« Training a specific neural network for each shape is neither feasible nor very useful. »

# Weight-encoded neural implicit



Ground Truth



Deep SDF reconstruction

« We propose training a specific neural network for each shape and will show that this approach is both feasible and very useful. »

# Weight-encoded neural implicit

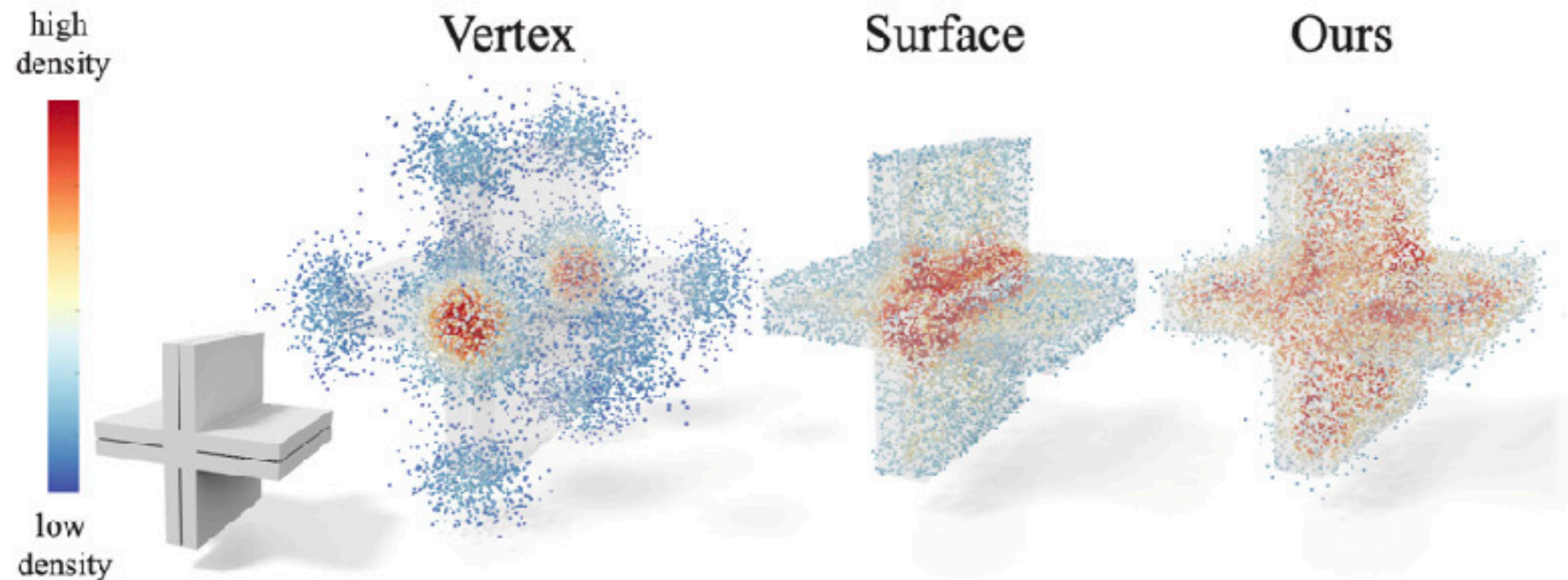
## Training:

- Sample  $N = 10^6$  points in the bounding volume
- Compute their true signed distance to the 3D closed mesh  $g_S(x)$
- Train a neural network with the  $L_1$  loss:  $|f_\theta(x) - g_S(x)|$
- Store the weights  $\theta$

## Inference:

- Load the estimated SDF  $f_\theta$
- Use it to render the shape (sphere tracing) or extract the surface (marching cubes)

# Weight-encoded neural implicit

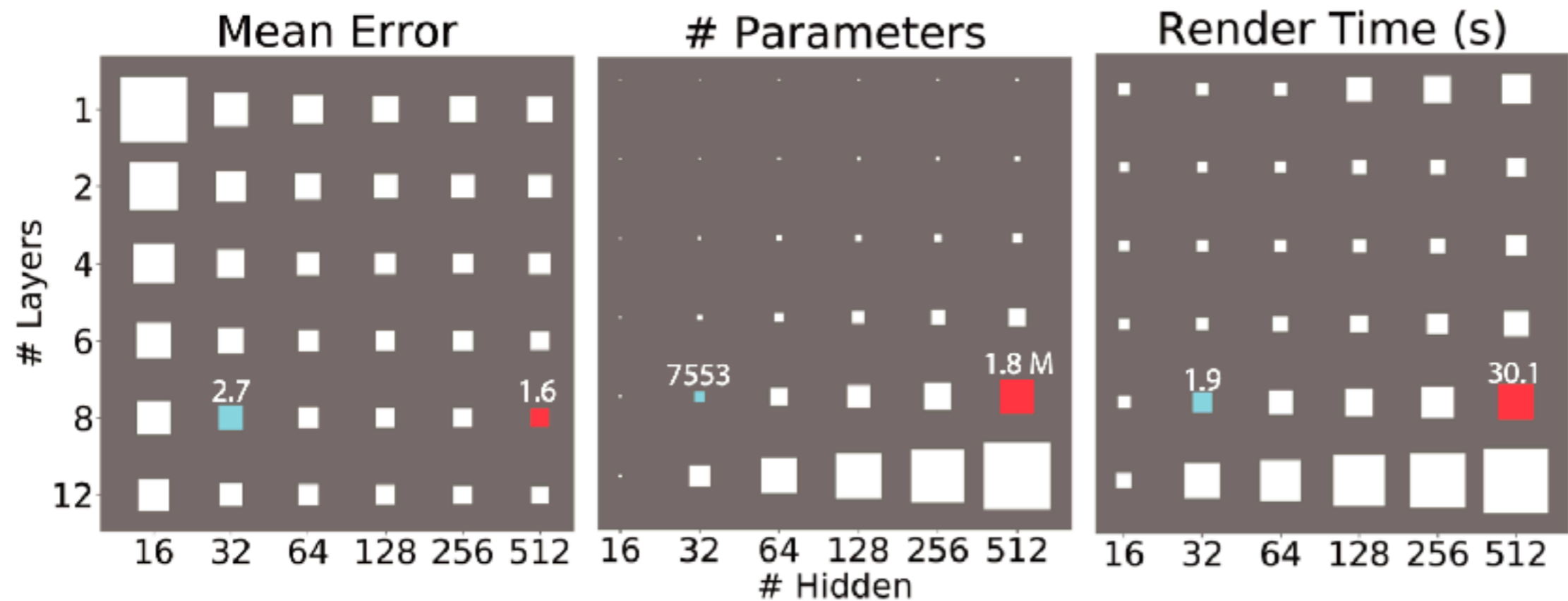


## Sampling the SDF

Uniform samples, discard w.r.t.  $1 - e^{-\beta|g_S(x)|}$

Source [Davies et al. 2020]

# Weight-encoded neural implicit



## Network architecture

7553 parameters: 8 layers, 32 features

59 kB

Source [Davies et al. 2020]

# Weight-encoded neural implicit



Ground Truth



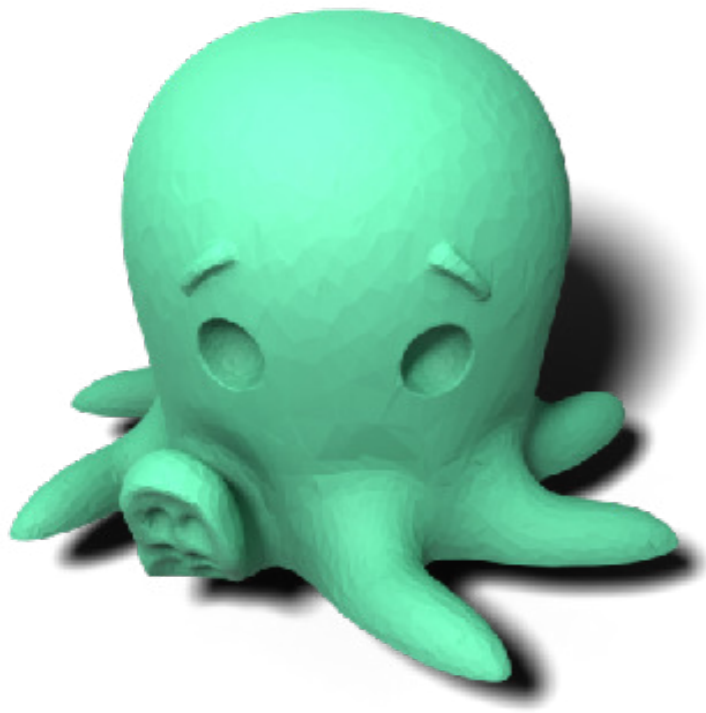
Neural Implicit



Quadric Error Metric

Sources [Davies et al. 2020] [Garland et al. 1997]

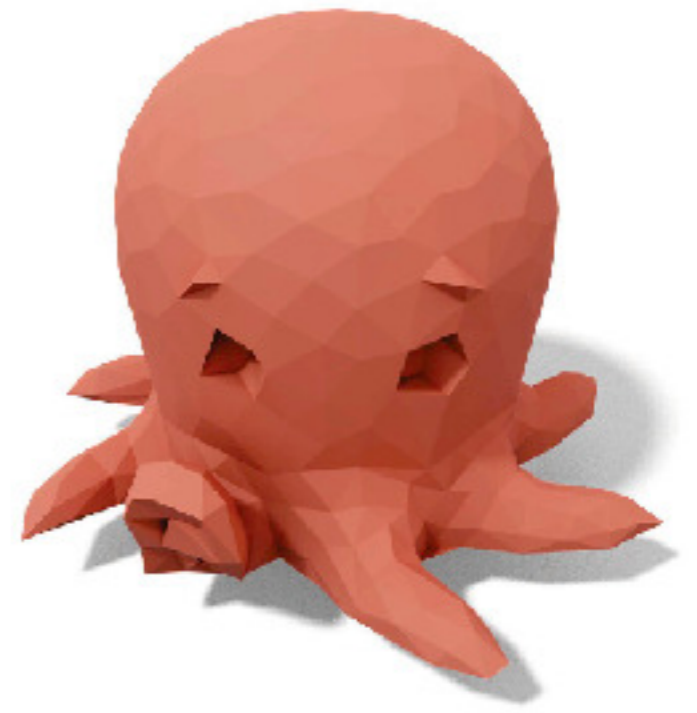
# Weight-encoded neural implicit



Ground Truth



Neural Implicit



Quadric Error Metric

Sources [Davies et al. 2020] [Garland et al. 1997]

# Conclusion

## Compact representation:

- 59 kB (66x size reduction)
- Relatively quick training (1 m) and inference time (1.9 s)
- Compared to classical approaches: no limits on the resolution

## Drawbacks:

- Heavy sampling
- Loss of details

# References

- [Davies et al. 2019] Davies, T., Nowrouzezahrai, D., & Jacobson, A. (2020). *On the effectiveness of weight-encoded neural implicit 3D shapes*. arXiv preprint arXiv:2009.09808.
- [Garland et al. 1997] Garland, M., & Heckbert, P. S. (1997, August). *Surface simplification using quadric error metrics*. In Proceedings of the 24th annual conference on Computer graphics and interactive techniques (pp. 209-216).
- [Park et al. 2019] Park, J. J., Florence, P., Straub, J., Newcombe, R., & Lovegrove, S. (2019). *DeepSDF: Learning continuous signed distance functions for shape representation*. In Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition (pp. 165-174).