## Method - Generalizable Shape Network

We approximate the object's occupancy  $\sigma(x) \approx \sigma_S(x) := f_\theta(x \mid S)$  where  $f_\theta$  is a neural network given shape parameters.

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$$\text{single latent code and obtain the occupancy value} \xrightarrow{\delta_{x}} \begin{cases} q_{i}^{S}(x)\}_{i} \\ f_{\phi_{o}}^{O} \end{cases} \xrightarrow{\kappa} = \kappa(\max(0, d_{i}(x, C_{i})))$$

$$\kappa = \left(1 - \left(\frac{a_{i}}{\rho}\right)^{2}\right)^{3}$$

$$\sigma_{S}(x) := f_{\phi_{o}}^{o}\left(\sum_{i} w_{i}^{S}(x)q_{i}^{S}(x)\right)^{\text{work}} \xrightarrow{\text{occupancy network}} \text{occupancy network}$$

ho define the extent of the joint regions,  $\kappa$  provide a smooth falloff to 0 as  $a_i$  go to ho.