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Convex Quantization Preserves Logconcavity

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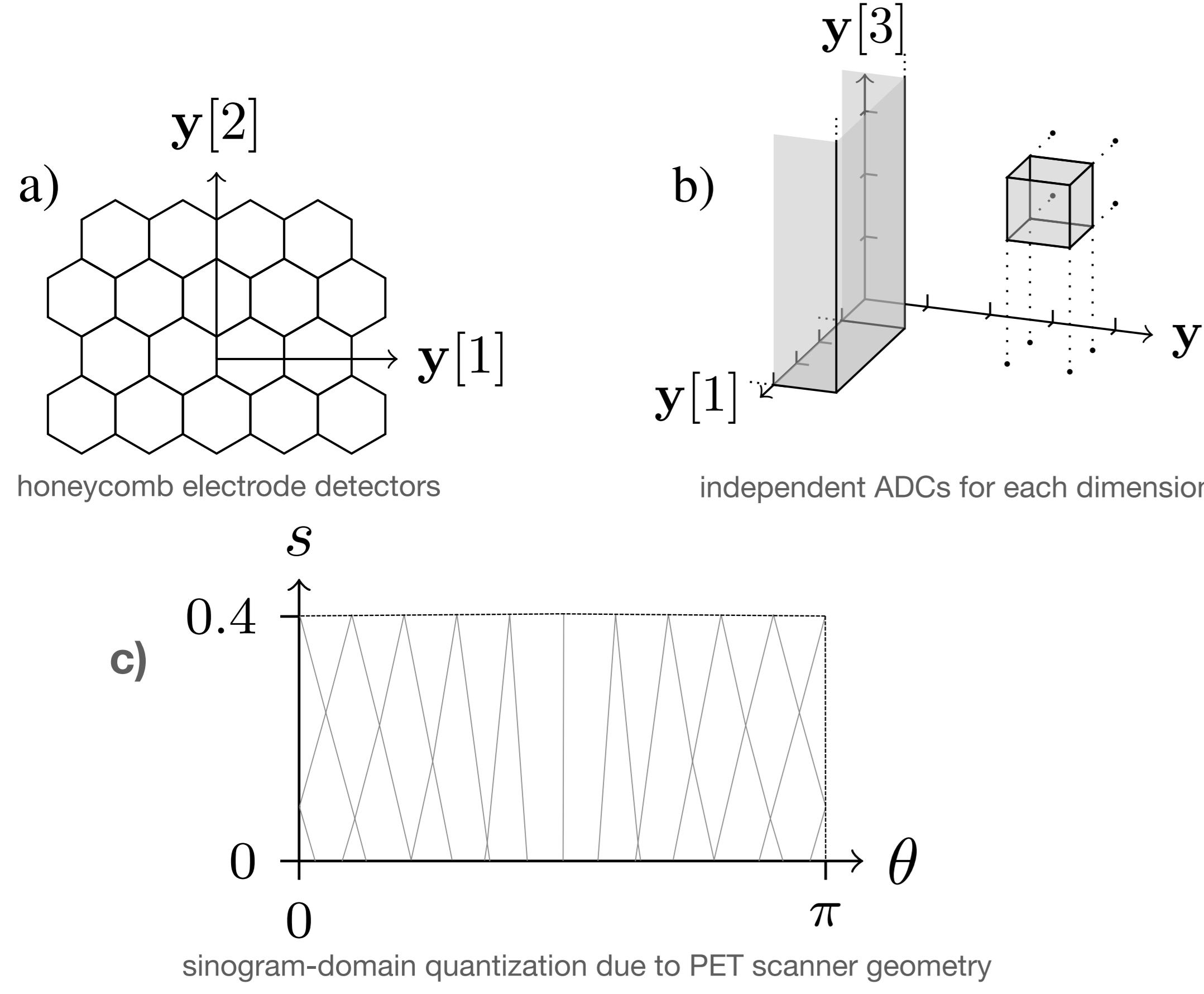
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Motivation

- Reality (quantization)

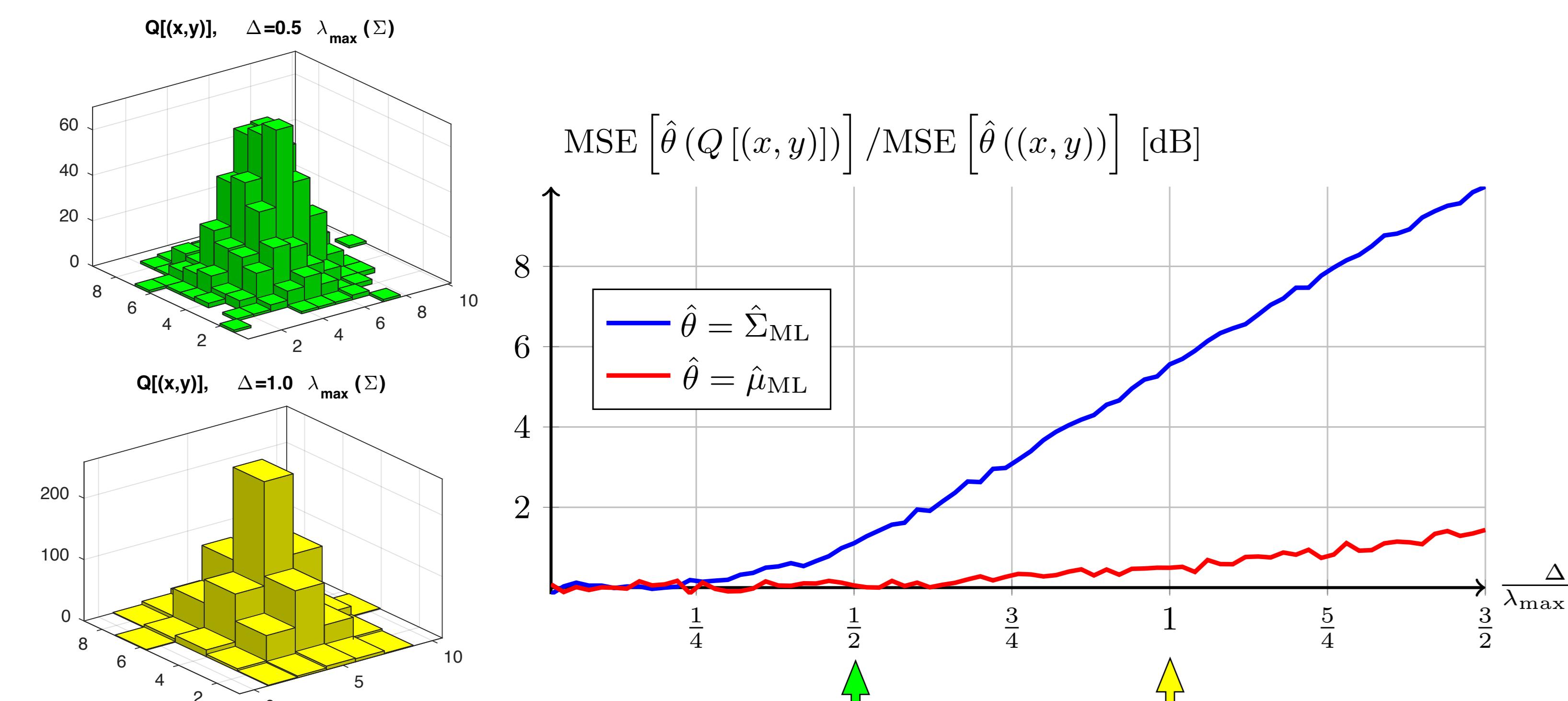


- Our models (continuous)

$$\mathbf{y} = \Psi^{-1} (\mathbf{S}\mathbf{x} + \mathbf{w}),$$

with $\Psi \in \mathcal{M}_n^+(\mathbb{R})$, $\mathbf{S} \in \mathcal{M}_{n,m}(\mathbb{R})$, and $\mathbf{x} \in \mathbb{R}^m$

- Why is that a problem? (wrong model)



Central Question

- Likelihood Logconcavity

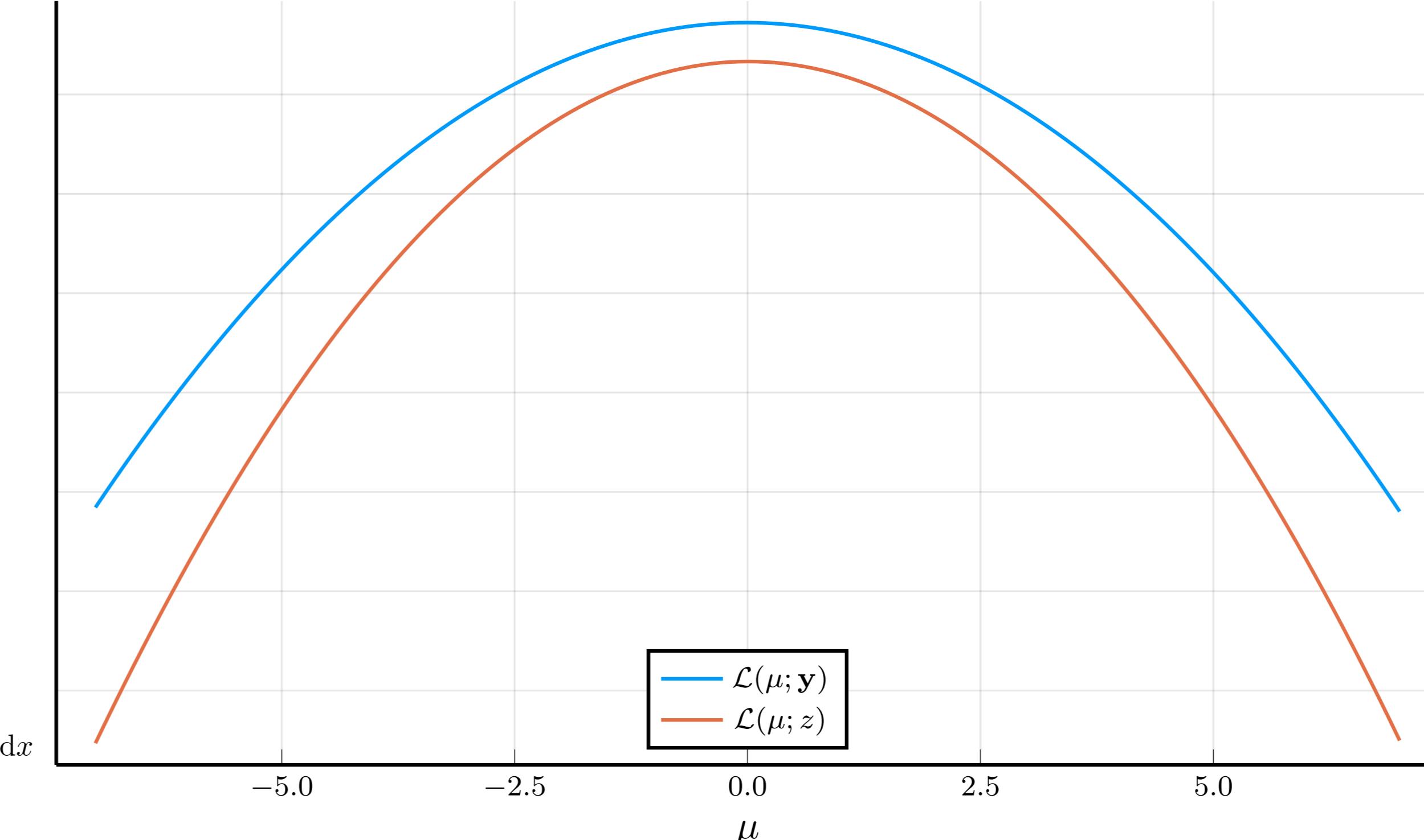
$$z = Q(\mathbf{y}) = Q(\Psi^{-1}(\mathbf{S}\mathbf{x} + \mathbf{w})), \quad (1)$$

 $Q: \mathbb{R}^n \rightarrow \mathcal{Z}$, where \mathcal{Z} is a countable set.

Is $\mathcal{L}(\mathbf{x}, \Psi; z)$ logconcave w.r.t. \mathbf{x} and Ψ ?

Example

- Standard normal samples
 $z_n = Q_\Delta(y_n)$
 $y_n \sim \mathcal{N}(0, \sigma)$
- Continuous-data likelihood
 $\mathbf{y} = [y_1, y_2, \dots, y_n]$
 $\mathcal{L}(\mu; \mathbf{y}) = \prod_n g_\sigma(y_n)$
- Quantized-data likelihood
 $z = (z_1, z_2, \dots, z_n)$
 $\mathcal{L}(\mu; z) = \prod_n \int_{z_n - \frac{\Delta}{2}}^{z_n + \frac{\Delta}{2}} g_\sigma(x) dx$



Our Answer

Same condition needed in cont. models

Theorem 1 **[Logconcave Noise and Convex Quantizers]** Consider a sample $z \in \mathcal{Z}$ drawn from model (1). Assume that Q is a convex quantizer and that the pdf $f_w(\cdot)$ of the noise is logconcave. Then,

- for a given scale parameter $\Psi_0 \in \mathcal{M}_n^+(\mathbb{R})$, the likelihood $\mathcal{L}(\mathbf{x}, \Psi_0; z)$ is logconcave with respect to \mathbf{x} ,
- for scale parameters of the form $\Psi = \psi \mathbf{I}$ with $\psi > 0$, the likelihood $\mathcal{L}(\mathbf{x}, \psi \mathbf{I}; z)$ is jointly logconcave with respect to \mathbf{x} and ψ ,
- for diagonal positive-definite scale parameters $\Psi = \Lambda$, the likelihood $\mathcal{L}(\mathbf{x}, \Lambda; z)$ is jointly logconcave with respect to \mathbf{x} and Λ if Q is composed of independent ADCs for each dimension.

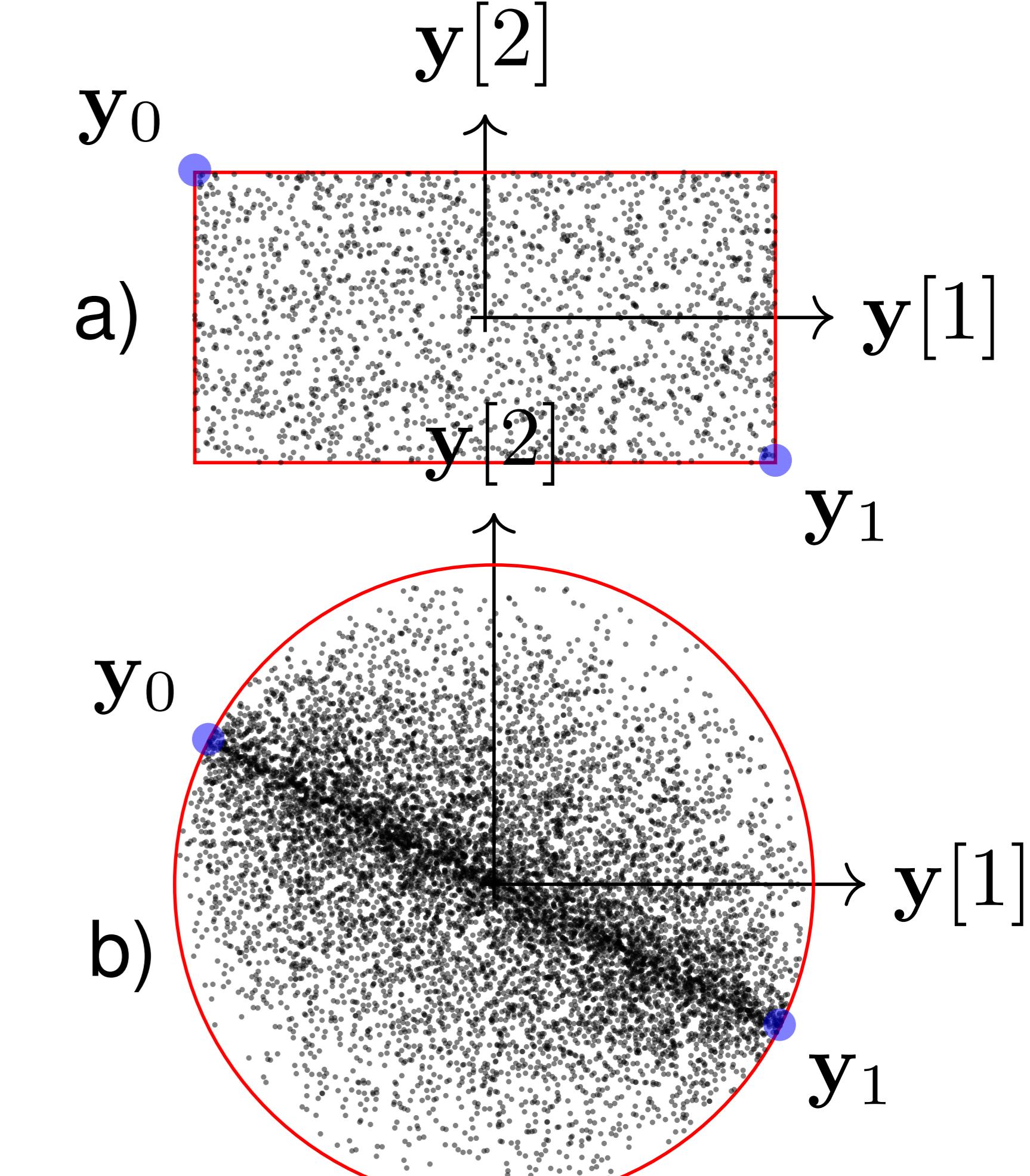
How?

- Study of the sets

$$\{\mathbf{y} : \mathbf{C}\mathbf{y}_1 + (\mathbf{I} - \mathbf{C})\mathbf{y}_0\}$$

when \mathbf{C} are

- random diagonal matrices
- random p.s.d. matrices

with $\rho(\mathbf{C}) \leq 1$.


Prékopa's Theorem

Logconcave probability laws on Minkowski averages of convex sets.

Key realisation

$$\mathcal{L}(\mathbf{x}, \Psi; z) = P_{\mathbf{w}} [\mathcal{W}_z(\mathbf{x}, \Psi)],$$

with $\mathcal{W}_z(\mathbf{x}, \Psi) = \{\mathbf{w} \in \mathbb{R}^n : \mathbf{y} \in Q^{-1}(z)\}$.