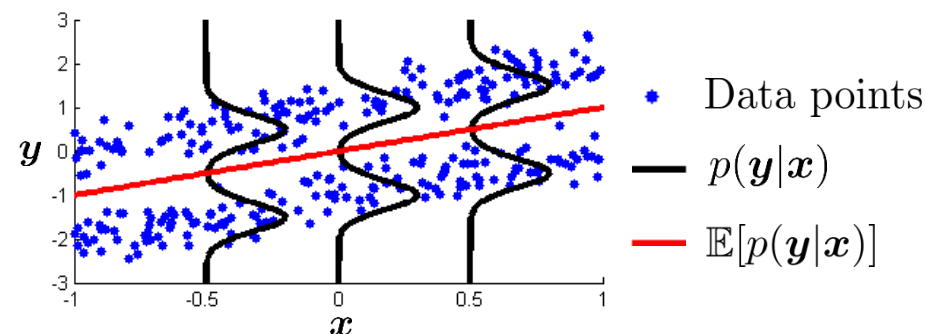


Conditional density estimation:

- In *regression*, we estimate the **mean** $\mathbb{E}[\cdot]$ of **conditional density** $p(\mathbf{y}|\mathbf{x})$.
 - The **conditional mean** $\mathbb{E}[p(\mathbf{y}|\mathbf{x})]$ alone is not informative enough to understand the data when conditional density possesses *multimodality*, *heteroscedasticity* and *asymmetry*.
 - We want to estimate the **conditional density** itself.



Existing method: Least-Squares Conditional Density Estimation (LSCDE)

- Least-squares fitting of conditional density model to true $p(\mathbf{y}|\mathbf{x})$.
- Optimal asymptotic convergence rate.
- The estimation is still challenging in **high dimensionality**.

Our contribution: Combine *conditional density estimation* with *dimensionality reduction*.

- We utilize **square-loss** variant of **conditional entropy**,

$$\text{SCE}(Y|X) = -\frac{1}{2} \int (p(\mathbf{y}|\mathbf{x}) - 1)^2 p(\mathbf{x}) d\mathbf{y}d\mathbf{x}.$$