## Normalizing Flows (Part 4)

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## **Linear Flows**

*	We will discuss different invertible network layers or flows, starting with linear flows.
<b>*</b>	Linear Flow Definition:
	$\square$ Form: $f[h] = \beta + \Omega h$
	$\square$ This is a bijection if and only if $\Omega \in \mathbb{R}^{D \times D}$ is an invertible square matrix.
*	Base Distribution Transformation:
	$\square$ Base distribution: Gaussian $P_r(z) = \mathcal{N}(\mu, \Sigma)$ $\square$ After linear transformation: $\mathcal{N}(\beta + \Omega  \mu, \Omega  \Sigma  \Omega^T)$
*	Expressiveness:
	☐ Linear flows alone are not sufficiently expressive. ☐ Useful when combined with nonlinear transformations.
*	The determinant of the Jacobian is simply the determinant of $\Omega$ .
*	We need to ensure that the Jacobian determinant and the inverse of the flow are fast to compute.
*	Computational Complexity:
	In general, computing $\det(\Omega)$ and $\det(\Omega^{-1})$ requires $O(D^3)$ time. If $\Omega$ is diagonal, the cost is $O(D)$ , but the elements of $h$ do not interact. If $\Omega$ is triangular, the Jacobian determinant is the product of its diagonal elements, taking $O(D)$ time. Inverting the flow requires solving the triangular system $\Omega  h = f[h] - \beta$ using back-substitution, which takes $O(D^2)$ time.