2. Now, we know  $A = \{\bar{a}_1, \bar{a}_2, \dots, \bar{a}_K\}$ . This gives a linear transformation of  $\bar{a}_n = A \bar{z}_n + \bar{z}_n$ ;  $\bar{z}_n \sim \mathcal{N}(o, \varphi)$  with a prior of  $p(\bar{z}_n) = \mathcal{N}(o, \varphi^{\dagger} \bar{z}_k)$ . and likelihood  $p(\bar{x}_n | \bar{z}_{n,A}) = \mathcal{N}(x_n | A \bar{z}_n + b, \psi)$ .

Using Bayes Theorem for gaussians,  $\rho(\Xi_n) = \mathcal{N}(\Xi_n \mid 0, p) \wedge (1) \text{ where } \Lambda = \mathcal{N}(X_n) + (1) \text{ where } L = \psi^{-1}$  and  $\rho(X_n \mid \Xi_n, A) = \mathcal{N}(X_n) + (1) \wedge (1)$ 

We senow that  $\rho\left(\Xi_{n} \mid \chi_{n}, A\right) = \mathcal{N}\left(\Xi_{n} \mid \Sigma\left\{A^{T} \downarrow \chi_{n} + \Lambda \cdot O\right\}, \Sigma\right); \Sigma = (\Lambda + A^{T} \downarrow A)^{-1}$   $= \mathcal{N}\left(\Xi_{n} \mid (\gamma \downarrow_{n} + A^{T} \downarrow_{n}^{T} A)^{-1} A^{T} \downarrow_{n}^{T} \gamma_{n}^{T} \gamma_{n}$ 

Since In only depends on in and not ik porken, therefore,  $p(\overline{\pm}_{n}|_{\mathbf{X}_{n},A}) = N(\overline{\pm}_{n}|_{([\gamma I_{k}+A^{T}\psi^{T}A)^{-1}Y^{T}X_{n})}, (\gamma I_{k} + A^{T}\psi^{T}A)^{-1})$