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**Algorithm 1** Kernel Conditional Independence Test

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Ensure the  $X, Y, Z$  are continuous random variables with domains  $x, y, z$ , respectively. Define a measurable, positive definite kernel  $K_x$  on  $x$  and denote the corresponding RKHS as  $H_x$ . Similarly define  $K_y, H_y, K_z, H_z$ .

$$\langle g, \sum_{Y|X} f \rangle = E_{XY}(f(X)g(Y)) - E_X(f(X))E_Y(g(Y))$$

$$\sum_{XY|Z} = \sum_{YX} - \sum_{YZ} \sum_{ZZ}^{-1} \sum_{ZX} \quad (1)$$

$$\sum_{XY|Z} = 0 \iff X \perp\!\!\!\perp Y|Z \quad (2)$$

$$\varepsilon'_{YZ} = \{g' | g' = g'(Y) - E(g'|Z), g' \in L_Y^2\} \quad (3)$$

$$\bar{f}(\ddot{X}) = f(\ddot{X}) - E(f|Z) = f(\ddot{X}) - h_f^*(Z) \quad (4)$$

$$T_{UI} = \frac{1}{n} \text{Tr}(K_x, K_y) - \text{Test Statistic}$$

It follows that

$$T_{UI} = \frac{1}{n} \text{Tr}(K_{\ddot{X}|Z}, K_{\ddot{Y}|Z}) \text{ for the conditional case}$$

Compare conditional test statistic to bootstrapped null-distribution based on eigenvalues of Kernels of shuffled samples

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