10/24/23

Outline

1) Testing with nuisance parameters

2) UMPU multivariate tests

3) Conditioning on null sufficient stat

Nuisance Parameters

Common setup: Extra unknown parameters which are not of direct interest

$\mathcal{P} = \{P_{\theta,\lambda} : (\theta, \lambda) \in \Omega\}$, $H_0 : \theta \in \Theta_0$ vs $H_1 : \theta \in \Theta_1$

θ parameter of interest

λ nuisance parameter

Issue: λ unknown but might affect type I error or power of a given test

Ex $X_1, \ldots, X_n \overset{iid}{\sim} N(\mu, \sigma^2)$ $\quad Y_1, \ldots, Y_m \overset{iid}{\sim} N(\nu, \sigma^2)$

μ, ν, σ^2 unknown

$H_0 : \mu = \nu$ vs $H_1 : \mu \neq \nu$

$\theta = \mu - \nu$ $\quad \lambda = (\mu + \nu, \sigma^2)$ or (μ, σ^2)

Ex $X_1 \sim \text{Binom}(n_1, \pi_1)$ $\quad X_2 \sim \text{Binom}(n_2, \pi_2)$

n_1, n_2 known $\Rightarrow$ not nuisance parameters

$H_0 : \pi_1 \leq \pi_2$ vs $H_1 : \pi_1 > \pi_2$

Multiparameter Exp. Families

Assume $X \sim p_{\theta,\lambda}(x) = e^{\theta' T(x) + \lambda' U(x) - A(\theta,\lambda)} h(x)$

$\theta \in \mathbb{R}^s$, $\lambda \in \mathbb{R}^r$, both unknown.

How to test $H_0 : \theta \in \Theta_0$ vs $H_1 : \theta \in \Theta_1$?

<u>Idea</u>: Condition on $U(X)$ to eliminate dep. on λ

1) <u>Sufficiency reduction</u>:

$$(T(X), U(X)) \sim q_{\theta,\lambda}(t,u)$$

$$= e^{\theta' t + \lambda' u - A(\theta,\lambda)} g(t,u)$$

(density wrt e.g. Lebesgue on $\mathbb{R}^{s+r}$)

2) <u>Condition on $U(X)$</u>:

$$q_\theta(t \mid u) = \frac{q_{\theta,\lambda}(t,u)}{\int q_{\theta,\lambda}(z,u)\,dz}$$

$$= \frac{e^{\theta' t + \cancel{\lambda' u} - \cancel{A(\theta,\lambda)}} g(t,u)}{\underbrace{\int e^{\theta' z + \cancel{\lambda' u} - \cancel{A(\theta,\lambda)}} g(z,u)\,dz}_{= e^{B_u(\theta)}}}$$

$$= e^{\theta' t - B_u(\theta)} g(t,u)$$

3) <u>Conditional test:</u>

Test $H_0: \theta \in \Theta_0$ vs. $H_1: \theta \in \Theta_1$ in

s-parameter model $\mathcal{Q}_u = \{ q_\theta(t|u) : \theta \in \Theta \}$

<u>Note</u> if $s=1$, this family has MLR in T

Even if $s>1$, we still have gotten rid of λ

<u>Theorem</u> (Informal)

<u>Theorem</u> Let $\mathcal{P}$ be full rank exp. fam. with densities $p_{\theta,\lambda}(x) = e^{\theta T(x) + \lambda' U(x) - A(\theta,\lambda)} h(x)$

$\theta \in \mathbb{R}$, $\lambda \in \mathbb{R}^r$, $(\theta,\lambda) \in \Omega$ open, θ_0 possible

a) To test $H_0: \theta \le \theta_0$ vs. $H_1: \theta > \theta_0$, there is a UMPU test $\phi^*(x) = \psi(T(x); U(x))$ where

$$\psi(t;u) = \begin{cases} 1 & t > c(u) \\ \gamma(u) & t = c(u) \\ 0 & t < c(u) \end{cases}$$

With $c(u)$, $\gamma(u)$ chosen to make

$$\mathbb{E}_{\theta_0}[\phi^*(X) \mid U(X) = u] = \alpha$$

b) To test $H_0: \theta = \theta_0$ vs. $H_1: \theta \ne \theta_0$ there is a UMPU test $\phi^*(x) = \psi(T(x); U(x))$ where

$$\psi(t;u) = \begin{cases} 1 & t < c_1(u) \text{ or } t > c_2(u) \\ \gamma_i(u) & t = c_i(u) \\ 0 & t \in (c_1(u), c_2(u)) \end{cases}$$

with $c_i(u)$, $\gamma_i(u)$ chosen to make

$$\mathbb{E}_{\theta_0}[\phi^*(X) \mid U(X) = u] = \alpha$$

$$\mathbb{E}_{\theta_0}[T(X)(\phi^*(X) - \alpha) \mid U(X) = u] = 0$$

[Note λ has disappeared from the problem.]

Ex: $X_i \overset{ind.}{\sim} \text{Pois}(\mu_i) \qquad i = 1, 2$

$H_0: \mu_1 \le \mu_2$ vs. $H_1: \mu_1 > \mu_2$

$$p_\mu(x) = \prod_{i=1}^{2} \frac{\mu_i^{x_i} e^{-\mu_i}}{x_i!}$$

$$= e^{x_1 \eta_1 + x_2 \eta_2 - (e^{\eta_1} + e^{\eta_2})} \frac{1}{x_1! x_2!}$$

(Where $\eta_i = \log \mu_i$. $H_0: \eta_1 \le \eta_2 \quad H_1: \eta_1 > \eta_2$)

$$= e^{\overbrace{x_1}^{T(x)} \cdot \overbrace{(\eta_1 - \eta_2)}^{\theta} + \overbrace{(x_1 + x_2)}^{u(x)} \cdot \overbrace{\eta_2}^{\lambda} - A(\eta)} \frac{1}{x_1! x_2!}$$

$H_0: \theta \le 0$ vs $H_1: \theta > 1$

Reject for <u>conditionally</u> large values of X_1, given $X_1 + X_2 = u$

$$\mathbb{P}_\theta(X_1 = x_1 \mid U = u) = e^{x_1 \theta + \cancel{u\lambda} - \cancel{A(\cdot)}} \cdot \frac{1}{x_1!(u - x_1)!} \Big/ \sum_{x_1 = 0}^{u} (\cdot)$$

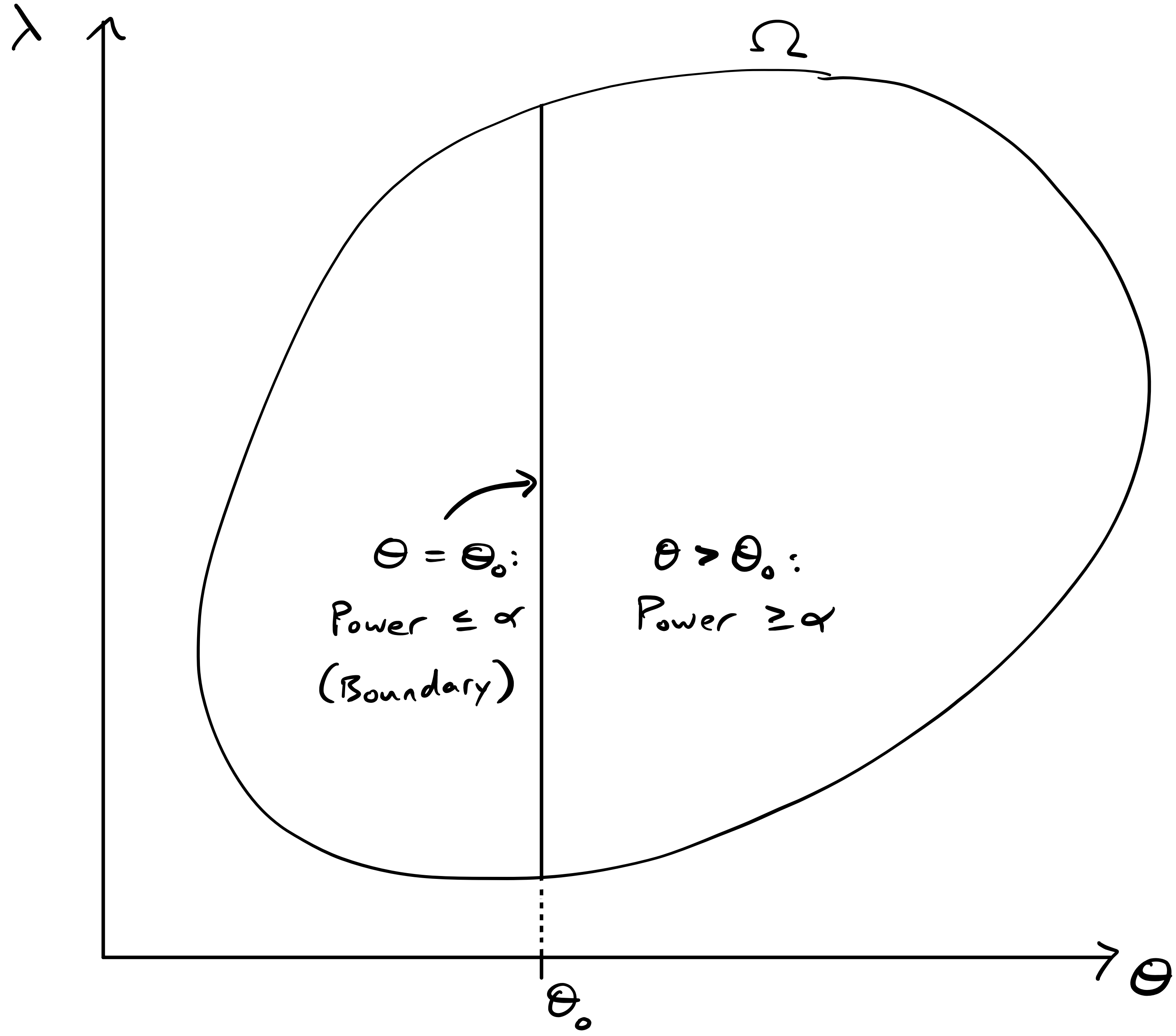
$$\propto_{x_1} e^{x_1 \theta} \cdot \frac{u!}{x!(u - x_1)!}$$

$$= \text{Binom}\left(u, \frac{e^\theta}{1 + e^\theta}\right) \qquad e^\theta = \mu_1 / \mu_2$$

$$= \text{Binom}\left(u, \frac{\mu_1}{\mu_1 + \mu_2}\right)$$

So in the end we do a Binomial test.

Proof Sketch



1) Any unbiased test has $\beta(\theta_0, \lambda) = \alpha \quad \forall \lambda$

(continuity)

2) Power $\equiv \alpha$ on boundary $\Rightarrow$ $\mathbb{E}_{\theta_0}[\phi \mid U] \overset{a.s.}{=} \alpha$

($U(X)$ complete sufficient on boundary submodel)

3) ϕ^* optimal among all tests with conditional level α

(by reduction to univariate model)

Proof Assume ϕ any unbiased test

Step 1: $\mathbb{E}_{\theta,\lambda}|\phi(X)| \leq 1 < \infty \quad \forall (\theta,\lambda) \in \Omega$

(Keener Thm 2.4)
$\Longrightarrow$ $\mathbb{E}_{\theta,\lambda}\phi(X)$ infinitely diff. on Ω, can diff. under $\int$

ϕ unbiased $\Rightarrow$ $\mathbb{E}_{\theta_0,\lambda}[\phi(X)] = \alpha \quad \forall (\theta_0,\lambda) \in \Omega$

Step 2: Boundary submodel: $\mathcal{P}_{\theta_0} = \{ P_{\theta_0,\lambda} : (\theta_0,\lambda) \in \Omega \}$

$$p_{\theta_0,\lambda}(x) = e^{\lambda' U(x) - A(\theta_0,\lambda)} \cdot \frac{e^{\theta_0 T(x)}}{h(x)}$$

$\mathcal{P}_{\theta_0}$ is full-rank, s-param exp. fam, $U(X)$ comp. suff.

Let $f(u) = \mathbb{E}_{\theta_0}[\phi(X) \mid U(X) = u] - \alpha$

$$\mathbb{E}_{\theta_0,\lambda}[f(U(X))] = \mathbb{E}_{\theta_0,\lambda}[\phi(X)] - \alpha = 0 \quad \forall \lambda$$

$$\Rightarrow f(u) \overset{a.s.}{=} 0$$

$$\Rightarrow \mathbb{E}_{\theta_0}[\phi(X) \mid U(x) = u] = 0 \quad \forall u$$

Two-sided case:
$$g(u) = \frac{d}{d\theta}\mathbb{E}_{\theta_0}[\phi \mid U = u]$$
$$= \mathbb{E}_{\theta_0}[(T - \mathbb{E}_{\theta_0}[T \mid u])\phi \mid u]$$
$$= \mathbb{E}_{\theta_0}[T(\phi - \alpha) \mid u]$$

$$\mathbb{E}_{\theta_0,\lambda}\, g(u) = \mathbb{E}_{\theta_0,\lambda}[T(\phi - \alpha)] = \frac{\partial}{\partial\theta}\beta_\phi(\theta_0) = 0 \quad \forall \lambda$$

$$\Longrightarrow \frac{d}{d\theta}\mathbb{E}_{\theta_0}[\phi \mid u] \overset{a.s.}{=} 0$$ (Cond'l power has derivative 0 at θ_0)

Step 3: For any value u, the conditional model is

$$q_\theta(t \mid u) = e^{\theta t - B_u(\theta)} g(t,u), \quad \text{1-param. exp. fam}$$

In one- / two-sided case, we have shown

$\psi(t;u)$ is UMP / UMPU in Q_u

Let $\bar{\phi}(t;u) = \mathbb{E}\left[\phi(X) \mid T(X)=t, U(X)=u\right]$

$$\mathbb{E}_\theta\left[\bar{\phi}(T;u) \mid U=u\right] = \mathbb{E}_\theta\left[\phi(X) \mid U(X)=u\right]$$
$$= \alpha \text{ if } \theta = \theta_0$$

$\Rightarrow \bar{\phi}(\cdot\,;u)$ is a (cond.'l) test of H_0 vs. H_1 in Q_u with power $= \alpha$ at boundary

One-sided case:

(or $\theta \le \theta_0$)

$\psi(t;u)$ is the UMP test of $\theta = \theta_0$ vs $\theta > \theta_0$ in Q_u, which is a 1-param. exp. fam.

Two-sided case:

$\psi(t;u)$ is the UMP test of $\theta = \theta_0$ vs. $\theta \ne \theta_0$ among tests with power $= \alpha$, $\frac{d}{d\theta}$power $= 0$ @ θ_0

(Keener Thm 12.22, main thm. for two-sided tests)

In either case ψ has higher cond. power than $\bar{\phi}$, a.s.

For $(\theta, \lambda) \in \Omega_1$:

$$\begin{aligned} \mathbb{E}_{\theta,\lambda}[\phi(X)] &= \mathbb{E}_{\theta,\lambda}\left[\mathbb{E}_{\theta}\left[\bar{\phi}(T; u) \mid u \right] \right] \\ &\leq \mathbb{E}_{\theta,\lambda}\left[\mathbb{E}_{\theta}\left[\psi(T; u) \mid u \right] \right] \\ &= \mathbb{E}_{\theta,\lambda}[\phi^*(X)] \end{aligned}$$

Ex $X_1, \dots, X_n \overset{iid}{\sim} N(\mu, \sigma^2)$ $\sigma^2 > 0$ unknown

$H_0: \mu = 0$ vs. $H_1: \mu \neq 0$

$$p_{\mu,\sigma^2}(x) = e^{\underbrace{\frac{\mu}{\sigma^2}}_{\theta} \underbrace{\sum X_i}_{T=\bar{X}} \underbrace{- \frac{1}{2\sigma^2}}_{\lambda} \underbrace{\sum X_i^2}_{u = \|X\|^2} - \frac{n\mu}{2\sigma^2}} \cdot \left(\frac{1}{2\pi\sigma^2}\right)^{n/2}$$

Optimal test rejects when $\bar{X}$ is extreme given $\|X\|$

If $\mu = 0$, p is rotationally symmetric

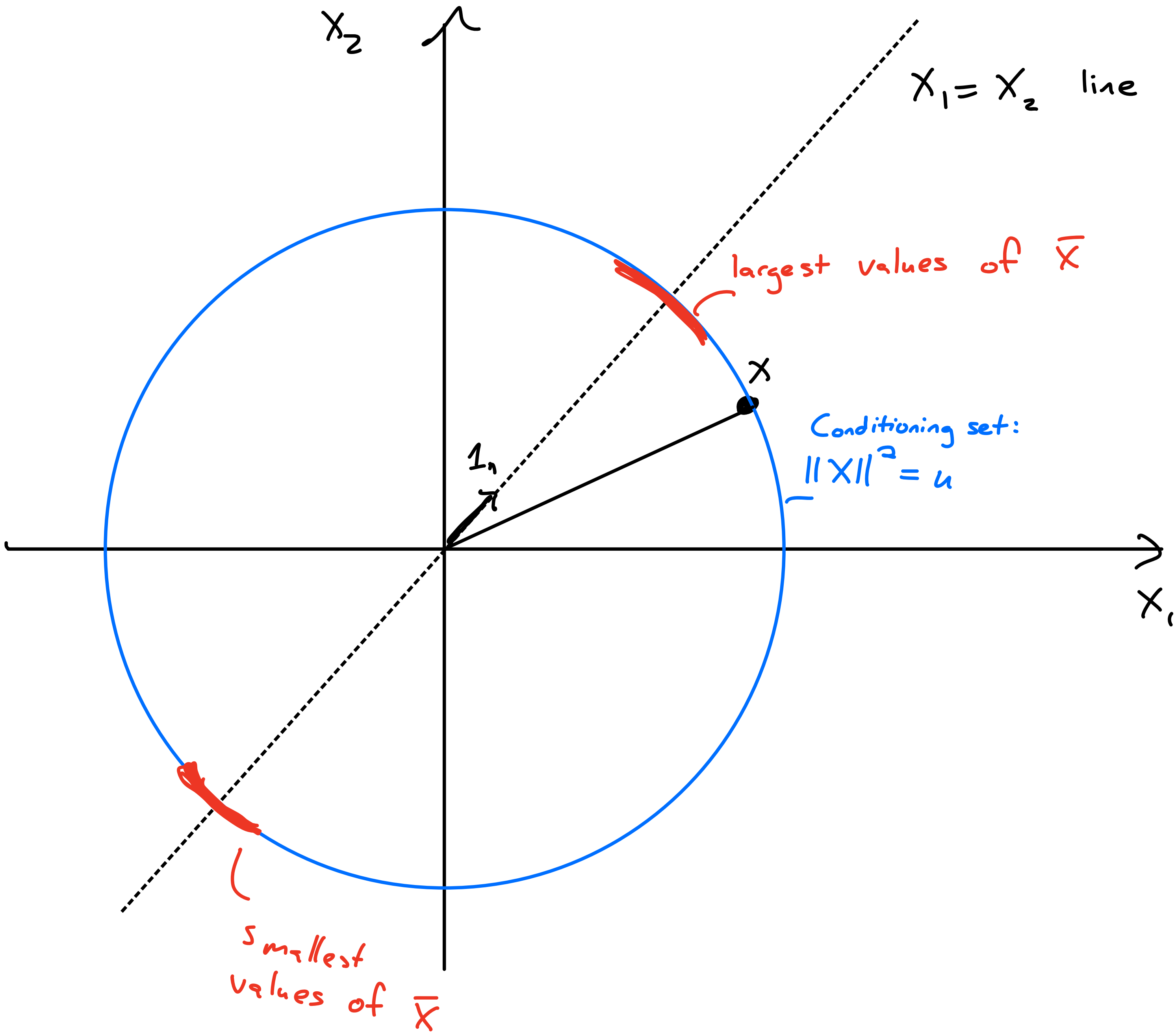
$$\Rightarrow X \,/\, \|X\|^2 = u \overset{H_0}{\sim} \text{Unif}(\sqrt{u} \cdot S^{n-1})$$

$$\left(\Leftrightarrow \frac{X}{\|X\|} \overset{H_0}{\sim} \text{Unif}(S^{n-1}), \text{ indep. of } \|X\| \right)$$

Optimal test rejects when $\frac{\bar{X}}{\|X\|}$ extreme (marginally)

Could stop here & simulate

Geometric Picture ($n=2$)



t - statistic

Above test rejects for

- conditionally extreme $\bar{X}$ given $\|X\|^2$

OR (equiv.)

- marginally extreme $\frac{\bar{X}}{\|X\|}$ $\left(\perp\!\!\!\perp \; \|X\|^2 \right)$

Equivalent: reject for marginally extreme

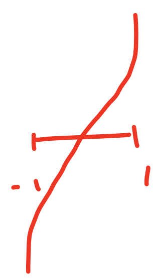
$$T = \frac{\sqrt{n}\,\bar{X}}{\sqrt{S^2}} \quad , \text{ where}$$

$$S^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2 \qquad \text{(sample variance)}$$

$$= \frac{1}{n-1} \left(\sum X_i^2 - 2\bar{X} \sum X_i + n\bar{X}^2 \right)$$

$$= \frac{1}{n-1} \left(\|X\|^2 - n\bar{X}^2 \right)$$

$r \to \frac{r}{\sqrt{1-r^2}}$:



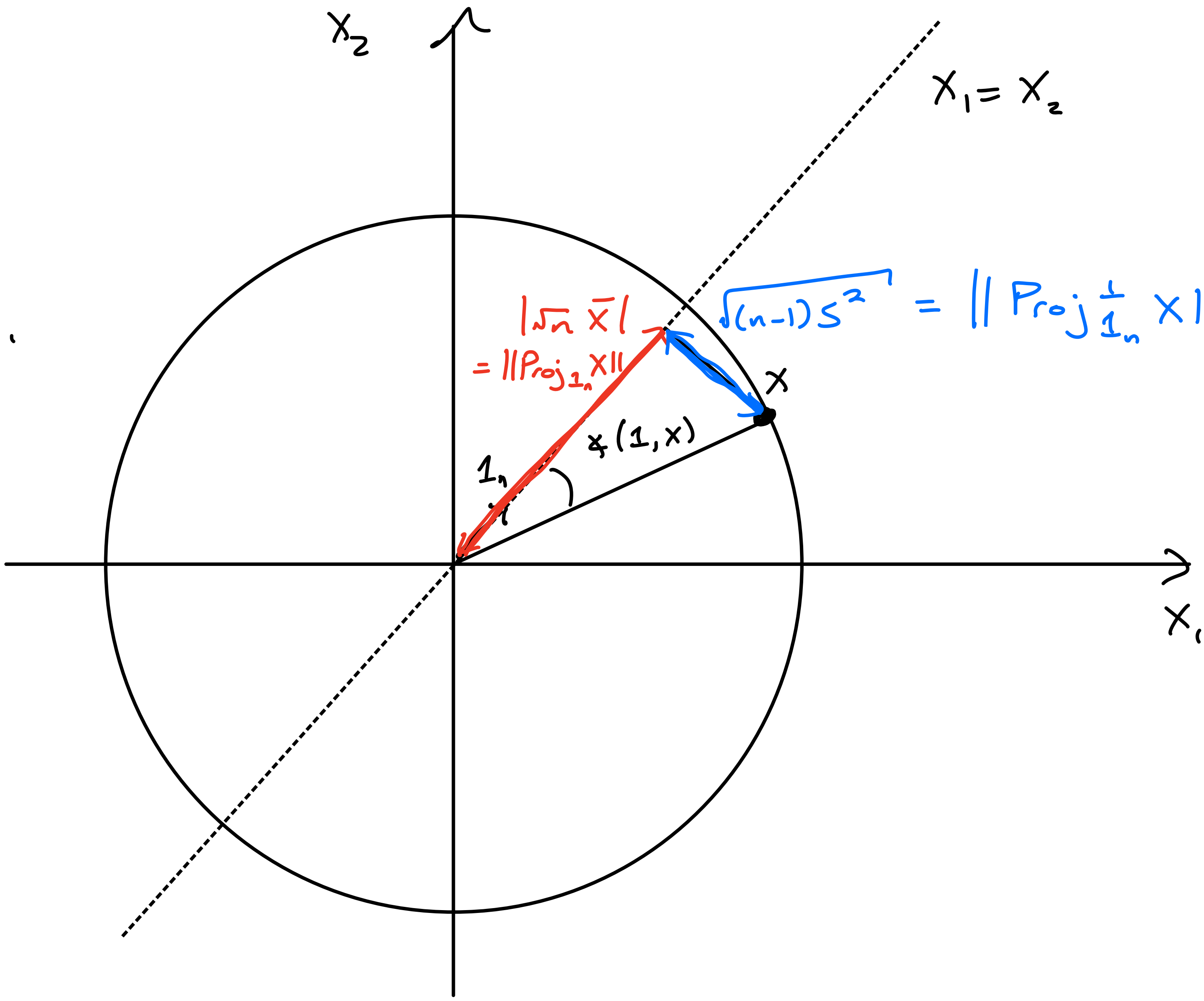
$$\Rightarrow \quad T = \sqrt{n-1} \cdot \frac{\sqrt{n}\,\bar{X}}{\sqrt{\|X\|^2 - n\bar{X}^2}} = \sqrt{n-1} \cdot \frac{R}{\sqrt{1-R^2}}$$

unit vectors

$$\text{for} \quad R = \frac{\sqrt{n}\,\bar{X}}{\|X\|} = \underbrace{\left(\frac{1}{\sqrt{n}} 1_n\right)' \frac{X}{\|X\|}}_{f\left(\frac{X}{\|X\|}\right) \Rightarrow \perp\!\!\!\perp \; \|X\|} = \cos \measuredangle (1_n, X)$$

Geometric Picture

$$T = \frac{\sqrt{n}\,\bar{X}}{\sqrt{S^2}} = \frac{\|\mathrm{Proj}_{1_n} X\|}{\|\mathrm{Proj}_{1_n}^{\perp} X\|} \cdot \sqrt{n-1}\,\mathrm{sgn}(\bar{X})$$



Next major theme: ratios of projections

Permutation Tests

Even if we don't get a UMPU test at the end, conditioning on null suff. stat. still helps.

Ex. $X_1, \dots, X_n \overset{iid}{\sim} P \quad Y_1, \dots, Y_m \overset{iid}{\sim} Q \quad H_0: P = Q \quad H_1: P \neq Q$

Under H_0, $P = Q$, $X_1, \dots, X_n, Y_1, \dots, Y_m \overset{iid}{\sim} P$

Let $(Z_1, \dots, Z_{n+m}) = (X_1, \dots, X_n, Y_1, \dots, Y_m)$

Under H_0, $U(Z) = (Z_{(1)}, \dots, Z_{(n+m)})$ compl. suff

Let $S_{n+m} = \{$Permutations on $n+m$ elements$\}$

$$(X, Y) \mid U \overset{H_0}{\sim} \text{Unif}(\{\pi U : \pi \in S_{n+m}\})$$

Thus, for <u>any</u> test stat T, if $P = Q$,

$$\mathbb{P}_{P,Q}(T(Z) \geq t \mid U) = \frac{1}{(n+m)!} \sum_{\pi \in S_{n+m}} 1\{T(\pi Z) \geq t\}$$

<u>Monte Carlo test</u>: In practice, we sample

$\pi_1, \dots, \pi_B \overset{iid}{\sim} S_{n+m}$, e.g. $B = 1000$

Then $Z, \pi_1 Z, \dots, \pi_B Z \overset{iid}{\sim} \text{Unif}(S_{n+m} U)$ under H_0

MC p-value $p = \frac{1}{1+B} \sum_{b=1}^{B} 1\{T(Z) \leq T(\pi_b Z)\}$

$\overset{H_0}{\sim} \text{Unif}(\{\frac{1}{1+B}, \dots, \frac{B-1}{1+B}, 1\})$ (if no ties)

($p \geq \text{Unif}(\cdot)$ if there are ties)