

(1)

Hypercontractivity:

Thm 4.12: Let $A: H_1 \rightarrow H_2$ be a linear operator between two Gaussian Hilbert Spaces, defined on probability spaces $(\Omega_i, \mathcal{F}_i, P_i)$ such that $\|A\| \leq 1$. Then $\Gamma(A)$ can be (uniquely) extended to a continuous linear operator $L^1(\Omega_1, \mathcal{F}(H_1), P_1) \rightarrow L^2(\Omega_2, \mathcal{F}(H_2), P_2)$, which we also denote by $\Gamma(A)$. Furthermore,

- i) $\|\Gamma(A)x\|_p \leq \|x\|_p$ for any $x \in L^p$, $1 \leq p \leq \infty$;
- ii) if $x \geq 0$ a.s., then $\Gamma(A)x \geq 0$ a.s.
- iii) $|\Gamma(A)x| \leq \Gamma(A)|x|$ a.s. for any $x \in L^1$.

Theorem 2. Let H_1 and H_2 be Gaussian Hilbert spaces defined on probability spaces $(\Omega_i, \mathcal{F}(H_i), P_i)$. If $1 \leq p \leq q \leq \infty$ and $A: H_1 \rightarrow H_2$ is a linear map with norm $\|A\| \leq (\frac{p-1}{q-1})^{\frac{1}{2}}$, then $\Gamma(A)$ maps $L^p(\Omega_1, \mathcal{F}(H_1), P_1)$ into $L^q(\Omega_2, \mathcal{F}(H_2), P_2)$ with norm $\|\Gamma(A)\|_{p,q} = 1$.

Remark: The condition $\|A\| \leq 1$ is sharp; if $\|A\| > \sqrt{(p-1)/(q-1)}$, then $\Gamma(A)$ does not even map L^p into L^q .

Reason: if $H_1 = H_2 \rightarrow$ one dimensional space. and $r^2 > (p-1)/(q-1)$.

Then $\Gamma(rI)(e^{ax^2}) \notin L^q$ for some a s.t. $e^{ax^2} \in L^p$.

$$\text{Note: } e^{ax^2} = \sum_{k=0}^{\infty} \frac{a^k}{k!} x^k = \sum_{k=0}^{\infty} \frac{a^k}{(k-1)! k!} x^{k-1} f_{k-1}(x) \Rightarrow \Gamma(rI)(e^{ax^2}) = r^2 e^{r^2 ax^2}.$$

Proof of thm 2: We have $\pi(A) = \pi_{\mathcal{P}(\|A\|_1)}(A/\|A\|_1)$. Hence it is suffice to consider the case $H_1 = H_2$ and $A = rI$ with $0 \leq r \leq \sqrt{\frac{p-1}{2r-1}}$.

Reason: $\pi(H) \xrightarrow{\pi(A/\|A\|_1)} \pi(H_2) \xrightarrow{\pi(A/\|A\|_1)} \pi(H_2)$

By Thm 1, we have

$$\|\pi(A)x\|_q = \left\| \pi_{\mathcal{P}(\|A\|_1)}(\pi(A/\|A\|_1)x) \right\|_q \stackrel{\substack{\text{show.} \\ ?}}{\leq} \|\pi(A/\|A\|_1)x\|_p \stackrel{\substack{\text{Thm 1} \\ ?}}{\leq} \|x\|_p.$$

claim 2: $\|\pi(rI)x\|_q \leq \|x\|_p$ for $x \in L^{\mathcal{P}(S_1, \mathcal{F}(H), P_1)}$.

$x = \{x_1, \dots, x_m\}$ → set of variables

Definition: $T_{x,r}(\sum_n y_n) = \sum_n r^n y_n$, $y_n \in \mathcal{P}_n(x) \cap \mathcal{P}_{n+1}(x)^\perp$

$\begin{cases} \text{if } y \in \mathcal{P}(x) \rightarrow \text{space of all polynomials generated by } x \\ y = \sum_n y_n \\ x = \{x_1, \dots, x_m\}. \end{cases}$

We say that the set $\{x_1, \dots, x_m\}$ is (B_2, r) -hypercontractive, if $T_{x,r}(y) \in \mathcal{P}(x)$ for all $y \in \mathcal{P}(x)$.

$$\|T_{x,r}(y)\|_q \leq \|y\|_p, \quad y \in \mathcal{P}(x).$$

Note, if the ~~the va~~ variables in x have a Gaussian distribution, then \mathcal{P} and \mathcal{P}_n are as defined in Ch 2. for Gaussian Hilbert space spanned by x and $T_{x,r} = \pi(rI)$. Hence we want to show any such set is (B_2, r) - hypercontractive.

Lemma 1. Let X be a random variable with the symmetric two point distribution $P(X=1) = P(X=-1) = \frac{1}{2}$. Then X is (p, q, r) -hypercontractive.

Lemma 2: Suppose that $X = \{x_1, \dots, x_m\}$ and $Y = \{y_1, \dots, y_n\}$ are two independent (p, q, r) -hypercontractive sets of random variables. Then their union $\{x_1, \dots, x_m, y_1, \dots, y_n\}$ is (p, q, r) -hypercontractive.

Proof of claim 1. Fix p, q , and r with $r^2 \leq \frac{p-1}{q-1}$. Let x_1, \dots, x_m be independent random variables with the symmetric distribution as in Lemma 1. And define $S_m = \sum_{i=1}^m X_i$. By Lemma 1 & Lemma 2, $\{x_1, \dots, x_m\}$ is (p, q, r) -hypercontractive.

claim 1: i) S_m is (p, q, r) -hypercontractive.

↓ ii) $\frac{S_m}{\sqrt{m}}$ is (p, q, r)

↓ iii) $\{\xi_i\}_{i=1}^m$ is (p, q, r)

↓ iv) $\{\xi_1, \xi_2, \dots, \xi_m\}$ is (p, q, r)

↓ v) $X \in P_n(H)$, $X = f(\xi_1, \dots, \xi_m)$ for some polynomial f and Gaussian variables $\xi_i \in H$, and $\Gamma(f)(x) = \mathbb{E}_{\xi_1, \dots, \xi_m} f(x)$.

Hence, $\|\Gamma(f)(x)\|_q \leq \|x\|_p$.

since, the set of such variables dense in L^p .

Hence, $\|\Gamma(f)(x)\|_q \leq \|x\|_p$ $\forall x \in L^p(S_m, J(H), P)$

Hence, the claim 1.

Proofs ii) It is clear that $P(S_m) \subseteq P(x_1, \dots, x_m)$.

$P_n(S_m) \rightarrow$ consists of all symmetric polynomials in $P_n(x_1, \dots, x_m)$

Thus if $x \in P_n(S_m)$ & $x \in P_n(S_m)^\perp \Rightarrow x \in P_n\{x_1, \dots, x_m\}^\perp$

$$i) P_n(S_m) \cap P_{n-1}(S_m)^\perp \subseteq P_n\{x_1, \dots, x_m\} \cap P_{n-1}\{x_1, \dots, x_m\}^\perp$$

and $T_{S_m, r}(y) = T_{\{x_1, \dots, x_m\}, r}(y) + y \in P(S_m)$.

$\Rightarrow S_m$ is (p, q, r) -hypercontractive.

ii) As, consequently, $\tilde{s}_m = \frac{s_m}{r_m}$ is (p, q, r) -hypercontractive.

iii) Now, $T_{\tilde{s}_m, r} \left(\sum_{k=0}^m a_k P_{m,k}(\tilde{s}_m) \right) = \sum_{k=0}^m a_k r^k P_{m,k}(\tilde{s}_m)$

Now let $m \rightarrow \infty$. $\tilde{s}_m \rightarrow \xi$ ($N(0, 1)$) [by CLT]

$P_{m,k}(\tilde{s}_m) \rightarrow h_k(\xi)$
↳ Hermit polynomials.

$P_{m,k}$, $k=0, \dots, m$
are orthogonal
polynomials for
the distribution of \tilde{s}_m .
upto change of
variable, these are
orthogonal polynomial
w.r.t. to B.D.
Krawtchouk polyn.

It follows that for any finite sequence a_0, \dots, a_l holds.

$$\sum_{k=0}^l a_k P_{m,k}(\tilde{s}_m) \rightarrow \sum_{k=0}^l a_k h_k(\xi) \text{ in } L^p \text{ distribution and in every } L^p.$$

consequently, using (p, q, r) -hypercontractivity of each \tilde{s}_m ,

$$\| T_{\xi, r} \left(\sum_k a_k h_k(\xi) \right) \|_q = \left\| \sum_k a_k r^k h_k(\xi) \right\|_q$$

$$= \lim_{m \rightarrow \infty} \left\| \sum_k a_k r^k P_{m,k}(\tilde{s}_m) \right\|_q$$

$$= \lim_{m \rightarrow \infty} \left\| \sum_k a_k T_{\tilde{s}_m, r}(P_{m,k}(\tilde{s}_m)) \right\|_q$$

$$\leq \lim_{m \rightarrow \infty} \left\| \sum_k a_k P_{m,k}(\tilde{s}_m) \right\|_p$$

$$= \left\| \sum_k a_k h_k(\xi) \right\|_p$$

$\Rightarrow \xi$ is (p, q, r) -hypercontractive.

Proof of Lemma 1. The space $P(X)$ of polynomial variables is two dimensional (3)

and spanned by 1 and X , with $T_{X,r} : a+bx \rightarrow a+bx^r$.

Hence, claim can be written

$$\left(\frac{|a+rb|^q + |a-rb|^q}{2} \right)^{\frac{1}{q}} \leq \left(\frac{|a+b|^p + |a-b|^p}{2} \right)^{\frac{1}{p}} \quad \begin{array}{l} \text{(this is known)} \\ \text{as two point} \\ \text{inequality} \end{array}$$

Note: $U: a+bx \rightarrow a-bx$, then $T_{X,r} = \frac{1+r}{2}(I+U) + \frac{1-r}{2}(I-U)$

$$\Rightarrow |T_{X,r}(f(x))| \leq \frac{1+r}{2}|f(x)| + \frac{1-r}{2}|U(f(x))| = T_{X,r}(|f(x)|).$$

Hence it is suffices to consider the case when $a+bx > 0$ i.e. $a > 0$

and $-a \leq b \leq a$.

~~(*)~~ we may assume $a=1$, $|b|<1$.

$$\therefore \left(\frac{(1+rb)^q + (1-rb)^q}{2} \right)^{\frac{1}{q}} \leq \left(\frac{(1+b)^p + (1-b)^p}{2} \right)^{\frac{1}{p}}$$

$$\Rightarrow \left(\sum_{k=0}^{\infty} \binom{q}{2k} r^{2k} b^{2k} \right)^{p/q} \leq \left(\sum_{k=0}^{\infty} \binom{p}{2k} b^{2k} \right)$$

For $1 \leq p \leq q \leq 2$.

$$\text{Let } L = \left(1 + \sum_{k=1}^{\infty} \binom{q}{2k} r^{2k} b^{2k} \right)^{p/q}$$

$$\leq 1 + \frac{p}{q} \sum_{k=1}^{\infty} \binom{q}{2k} r^{2k} b^{2k}$$

$$\leq 1 + \sum_{k=1}^{\infty} \binom{p}{2k} b^{2k}$$

$$= \sum_{k=0}^{\infty} \binom{p}{2k} b^{2k}$$

$$\because (1+x)^q \leq 1+qx \quad \text{for } x \geq 0, \quad 0 \leq x \leq 1$$

$$\therefore \binom{p}{q} \binom{q}{2k} r^{2k} \leq \binom{p}{2k}$$

$$\therefore \frac{p}{q} \cdot \frac{q(q-1)\dots(q-2k+1)}{p(p-1)\dots(p-2k+1)} \cdot \frac{p-2k}{(p-1)k}$$

$$\leq \frac{(q-1)\dots(q-2k+1)}{(p-2)\dots(p-2k+1)} \leq 1.$$

Lemma 1 is true for $1 \leq p \leq q \leq 2$. By duality we obtain the result for $0 \leq q \leq p \leq \infty$.

for $p \leq q$ the following for the $p \leq q \leq 2$ and $2 \leq p \leq q$

$$\frac{1}{1+p} \frac{1}{1+q} \frac{1}{2}$$

Proof of lemma 2: Any polynomial variable in $P(x \cup y)$ can be written as a finite sum $\sum_i f_i(x)g_i(y)$, where f_i, g_i are polynomials, and

$$T_{x,y,r} \left(\sum_i f_i(x)g_i(y) \right) = \sum_i \left(T_{x,r} f_i(x) \right) T_{y,r} g_i(y)$$

Let M_x and M_y denote distribution of x and y .

Now,

$$\begin{aligned} \left\| \sum_i T_{x,r} f_i(x) T_{y,r} g_i(y) \right\|_q &= \left\| \left\| \sum_i T_{y,r} \left(\sum_i T_{x,r} f_i(x) g_i(y) \right) \right\|_{L^q(M_y)} \right\|_{L^q(M_x)} \\ &\stackrel{\text{contractivity}}{\leq} \left\| \left\| \sum_i T_{x,r} f_i(x) g_i(y) \right\|_{L^2(M_x)} \right\|_{L^2(M_y)} \\ \textcircled{c.4} &\leq \left\| \left\| \sum_i T_{x,r} f_i(x) g_i(y) \right\|_{L^2(M_x)} \right\|_{L^p(M_y)} \\ &\stackrel{\text{contractivity}}{\leq} \left\| \left\| \sum_i f_i(x) g_i(y) \right\|_{L^p(M_x)} \right\|_{L^p(M_y)} \\ &= \left\| \sum_i f_i(x) g_i(y) \right\|_{L^p}. \end{aligned}$$

(c.4) Let, (M_1, M_1, μ_1) and (M_2, M_2, μ_2) be two σ -finite measure spaces. If f is a measurable function on $M_1 \times M_2$ and $0 < p \leq q \leq \infty$, then

$$\left\| \|f\|_{L^p(\mu_1, M_1)} \right\|_{L^q(M_2, \mu_2)} \leq \left\| \|f\|_{L^2(\mu_2, M_2)} \right\|_{L^p(M_1, \mu_1)}$$