

Concentration Inequalities

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Abstract

Concentration inequalities provide bounds on how a random variable (or a sum/composition of random variables) deviate from their expectation, usually based on moments/independence of the variables.

The most important concentration inequalities (the Markov, Chebyshev, and Hoelder inequalities and the Chernoff bounds) are already part of HOL-Probability. This entry collects more advanced results, such as Bennett's/Bernstein's Inequality, Bienaymé's Identity, Cantelli's Inequality, the Efron-Stein Inequality, McDiarmid's Inequality, and the Paley-Zygmund Inequality.

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1 Preliminary results

```
theory Concentration-Inequalities-Preliminary
  imports Lp.Lp
begin
```

Version of Cauchy-Schwartz for the Lebesgue integral:

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lemma *cauchy-schwartz*:

fixes $f\ g :: - \Rightarrow \text{real}$
assumes $f \in \text{borel-measurable } M\ g \in \text{borel-measurable } M$
assumes $\text{integrable } M (\lambda x. (f\ x)^2)\ \text{integrable } M (\lambda x. (g\ x)^2)$
shows $\text{integrable } M (\lambda x. f\ x * g\ x)$ (**is** ?A)
 $(\int x. f\ x * g\ x\ \partial M) \leq (\int x. (f\ x)^2\ \partial M)^{\text{powr } (1/2)} * (\int x. (g\ x)^2\ \partial M)^{\text{powr } (1/2)}$
(**is** ?L \leq ?R)
{proof}

Generalization of *prob-space.indep-vars-iff-distr-eq-PiM'*:

lemma (**in** *prob-space*) *indep-vars-iff-distr-eq-PiM''*:

fixes $I :: 'i \text{ set}$ **and** $X :: 'i \Rightarrow 'a \Rightarrow 'b$
assumes $rv: \bigwedge i. i \in I \Longrightarrow \text{random-variable } (M' i) (X i)$
shows $\text{indep-vars } M' X I \longleftrightarrow$
 $\text{distr } M (\prod_M i \in I. M' i) (\lambda x. \lambda i \in I. X i x) = (\prod_M i \in I. \text{distr } M (M' i) (X i))$
{proof}

lemma *proj-indep*:

assumes $\bigwedge i. i \in I \Longrightarrow \text{prob-space } (M i)$
shows $\text{prob-space.indep-vars } (PiM\ I\ M) M (\lambda i\ \omega. \omega\ i) I$
{proof}

lemma *forall-Pi-to-PiE*:

assumes $\bigwedge x. P\ x = P (\text{restrict } x\ I)$
shows $(\forall x \in Pi\ I\ A. P\ x) = (\forall x \in PiE\ I\ A. P\ x)$
{proof}

lemma *PiE-reindex*:

assumes $\text{inj-on } f\ I$
shows $PiE\ I (A \circ f) = (\lambda a. \text{restrict } (a \circ f)\ I) \text{ ' } PiE (f \text{ ' } I) A$ (**is** ?lhs = ?g \text{ ' } ?rhs)
{proof}

context *prob-space*

begin

lemma *indep-sets-reindex*:

assumes $\text{inj-on } f\ I$
shows $\text{indep-sets } A (f \text{ ' } I) = \text{indep-sets } (\lambda i. A (f\ i)) I$
{proof}

lemma *indep-vars-reindex*:

assumes $\text{inj-on } f\ I$
assumes $\text{indep-vars } M' X' (f \text{ ' } I)$
shows $\text{indep-vars } (M' \circ f) (\lambda k\ \omega. X' (f\ k)\ \omega) I$
{proof}

lemma *indep-vars-cong-AE*:
assumes $AE\ x\ in\ M. (\forall i \in I. X' i\ x = Y' i\ x)$
assumes *indep-vars* $M' X' I$
assumes $\bigwedge i. i \in I \implies random\ variable\ (M' i)\ (Y' i)$
shows *indep-vars* $M' Y' I$
 $\langle proof \rangle$

end

Integrability of bounded functions on finite measure spaces:

lemma *bounded-const*: *bounded* $((\lambda x. (c::real))\ 'T)$
 $\langle proof \rangle$

lemma *bounded-exp*:
fixes $f :: 'a \Rightarrow real$
assumes *bounded* $((\lambda x. f\ x)\ 'T)$
shows *bounded* $((\lambda x. exp\ (f\ x))\ 'T)$
 $\langle proof \rangle$

lemma *bounded-mult-comp*:
fixes $f :: 'a \Rightarrow real$
assumes *bounded* $(f\ 'T)$ *bounded* $(g\ 'T)$
shows *bounded* $((\lambda x. (f\ x) * (g\ x))\ 'T)$
 $\langle proof \rangle$

lemma *bounded-sum*:
fixes $f :: 'i \Rightarrow 'a \Rightarrow real$
assumes *finite* I
assumes $\bigwedge i. i \in I \implies bounded\ (f\ i\ 'T)$
shows *bounded* $((\lambda x. (\sum i \in I. f\ i\ x))\ 'T)$
 $\langle proof \rangle$

lemma (*in finite-measure*) *bounded-int*:
fixes $f :: 'i \Rightarrow 'a \Rightarrow real$
assumes *bounded* $((\lambda x. f\ (fst\ x)\ (snd\ x))\ '(T \times space\ M))$
shows *bounded* $((\lambda x. (\int \omega. (f\ x\ \omega)\ \partial M))\ 'T)$
 $\langle proof \rangle$

lemmas *bounded-intros* =
bounded-minus-comp bounded-plus-comp bounded-mult-comp bounded-sum
finite-measure.bounded-int
bounded-const bounded-exp

lemma (*in prob-space*) *integrable-bounded*:
fixes $f :: - \Rightarrow ('b :: \{banach, second-countable-topology\})$
assumes *bounded* $(f\ 'space\ M)$
assumes $f \in M \rightarrow_M\ borel$
shows *integrable* $M\ f$
 $\langle proof \rangle$

end

2 Bennett's Inequality

In this section we verify Bennett's inequality [1] and a (weak) version of Bernstein's inequality as a corollary. Both inequalities give concentration bounds for sums of independent random variables. The statement and proofs follow a summary paper by Boucheron et al. [2].

theory *Bennett-Inequality*

imports *Concentration-Inequalities-Preliminary*

begin

context *prob-space*

begin

lemma *indep-vars-Chernoff-ineq-ge:*

assumes *I: finite I*

assumes *ind: indep-vars (λ -. borel) X I*

assumes *sge: $s \geq 0$*

assumes *int: $\bigwedge i. i \in I \implies$ integrable $M (\lambda x. \exp (s * X i x))$*

shows *prob { $x \in$ space $M. (\sum i \in I. X i x - \text{expectation } (X i)) \geq t$ } \leq*

*$\exp (-s*t) *$*

*$(\prod i \in I. \text{expectation } (\lambda x. \exp(s * (X i x - \text{expectation } (X i))))$*

\langle proof \rangle

definition *bennett-h::real \implies real*

where *bennett-h $u = (1 + u) * \ln (1 + u) - u$*

lemma *exp-sub-two-terms-eq:*

fixes *x :: real*

shows *$\exp x - x - 1 = (\sum n. x^{(n+2)} / \text{fact } (n+2))$*

$\text{summable } (\lambda n. x^{(n+2)} / \text{fact } (n+2))$

\langle proof \rangle

lemma *psi-mono:*

defines *$f \equiv (\lambda x. (\exp x - x - 1) - x^2 / 2)$*

assumes *xy: $a \leq (b::real)$*

shows *$f a \leq f b$*

\langle proof \rangle

lemma *psi-inequality:*

assumes *le: $x \leq (y::real)$ $y \geq 0$*

shows *$y^2 * (\exp x - x - 1) \leq x^2 * (\exp y - y - 1)$*

⟨proof⟩

lemma *bennett-inequality-1*:

assumes I : finite I

assumes ind : indep-vars $(\lambda -. borel) X I$

assumes $intsq$: $\bigwedge i. i \in I \implies integrable M (\lambda x. (X i x)^2)$

assumes bn d: $\bigwedge i. i \in I \implies AE x in M. X i x \leq 1$

assumes t : $t \geq 0$

defines $V \equiv (\sum i \in I. expectation (\lambda x. X i x^2))$

shows $prob \{x \in space M. (\sum i \in I. X i x - expectation (X i)) \geq t\} \leq$

$exp (-V * bennett-h (t / V))$

⟨proof⟩

lemma *real-AE-le-sum*:

assumes $\bigwedge i. i \in I \implies AE x in M. f i x \leq (g i x::real)$

shows $AE x in M. (\sum i \in I. f i x) \leq (\sum i \in I. g i x)$

⟨proof⟩

lemma *real-AE-eq-sum*:

assumes $\bigwedge i. i \in I \implies AE x in M. f i x = (g i x::real)$

shows $AE x in M. (\sum i \in I. f i x) = (\sum i \in I. g i x)$

⟨proof⟩

theorem *bennett-inequality*:

assumes I : finite I

assumes ind : indep-vars $(\lambda -. borel) X I$

assumes $intsq$: $\bigwedge i. i \in I \implies integrable M (\lambda x. (X i x)^2)$

assumes bn d: $\bigwedge i. i \in I \implies AE x in M. X i x \leq B$

assumes t : $t \geq 0$

assumes B : $B > 0$

defines $V \equiv (\sum i \in I. expectation (\lambda x. X i x^2))$

shows $prob \{x \in space M. (\sum i \in I. X i x - expectation (X i)) \geq t\} \leq$

$exp (- V / B^2 * bennett-h (t * B / V))$

⟨proof⟩

lemma *bennett-h-bernstein-bound*:

assumes $x \geq 0$

shows $bennett-h x \geq x^2 / (2 * (1 + x / 3))$

⟨proof⟩

lemma *sum-sq-exp-eq-zero-imp-zero*:

assumes finite I $i \in I$

assumes $intsq$: integrable $M (\lambda x. (X i x)^2)$

assumes $(\sum i \in I. expectation (\lambda x. X i x^2)) = 0$

shows $AE\ x\ in\ M.\ X\ i\ x = (0::real)$
 ⟨proof⟩

corollary *bernstein-inequality*:

assumes I : *finite* I
assumes ind : *indep-vars* $(\lambda\ -.,\ borel)\ X\ I$
assumes $intsq$: $\bigwedge i.\ i \in I \implies integrable\ M\ (\lambda x.\ (X\ i\ x)^2)$
assumes bn d: $\bigwedge i.\ i \in I \implies AE\ x\ in\ M.\ X\ i\ x \leq B$
assumes t : $t \geq 0$
assumes B : $B > 0$
defines $V \equiv (\sum i \in I.\ expectation\ (\lambda x.\ X\ i\ x^2))$
shows $prob\ \{x \in space\ M.\ (\sum i \in I.\ X\ i\ x - expectation\ (X\ i)) \geq t\} \leq$
 $exp\ (- (t^2 / (2 * (V + t * B / 3))))$
 ⟨proof⟩

end

end

3 Bienaymé's identity

Bienaymé's identity [5, §17] can be used to deduce the variance of a sum of random variables, if their co-variance is known. A common use-case of the identity is the computation of the variance of the mean of pair-wise independent variables.

theory *Bienaymes-Identity*

imports *Concentration-Inequalities-Preliminary*

begin

context *prob-space*

begin

lemma *variance-divide*:

fixes $f :: 'a \Rightarrow real$
assumes *integrable* $M\ f$
shows $variance\ (\lambda\ \omega.\ f\ \omega / r) = variance\ f / r^2$
 ⟨proof⟩

definition *covariance* **where**

$covariance\ f\ g = expectation\ (\lambda\ \omega.\ (f\ \omega - expectation\ f) * (g\ \omega - expectation\ g))$

lemma *covariance-eq*:

fixes $f :: 'a \Rightarrow real$
assumes $f \in borel\ measurable\ M\ g \in borel\ measurable\ M$
assumes *integrable* $M\ (\lambda\ \omega.\ f\ \omega^2)$ *integrable* $M\ (\lambda\ \omega.\ g\ \omega^2)$
shows $covariance\ f\ g = expectation\ (\lambda\ \omega.\ f\ \omega * g\ \omega) - expectation\ f\ expectation\ g$

* *expectation g*
<proof>

lemma covar-integrable:

fixes $f g :: 'a \Rightarrow \text{real}$
assumes $f \in \text{borel-measurable } M$ $g \in \text{borel-measurable } M$
assumes $\text{integrable } M (\lambda \omega. f \omega^2)$ $\text{integrable } M (\lambda \omega. g \omega^2)$
shows $\text{integrable } M (\lambda \omega. (f \omega - \text{expectation } f) * (g \omega - \text{expectation } g))$
<proof>

lemma sum-square-int:

fixes $f :: 'b \Rightarrow 'a \Rightarrow \text{real}$
assumes $\text{finite } I$
assumes $\bigwedge i. i \in I \Rightarrow f i \in \text{borel-measurable } M$
assumes $\bigwedge i. i \in I \Rightarrow \text{integrable } M (\lambda \omega. f i \omega^2)$
shows $\text{integrable } M (\lambda \omega. (\sum i \in I. f i \omega)^2)$
<proof>

theorem bienaymes-identity:

fixes $f :: 'b \Rightarrow 'a \Rightarrow \text{real}$
assumes $\text{finite } I$
assumes $\bigwedge i. i \in I \Rightarrow f i \in \text{borel-measurable } M$
assumes $\bigwedge i. i \in I \Rightarrow \text{integrable } M (\lambda \omega. f i \omega^2)$
shows
 $\text{variance } (\lambda \omega. (\sum i \in I. f i \omega)) = (\sum i \in I. (\sum j \in I. \text{covariance } (f i) (f j)))$
<proof>

lemma covar-self-eq:

fixes $f :: 'a \Rightarrow \text{real}$
shows $\text{covariance } f f = \text{variance } f$
<proof>

lemma covar-indep-eq-zero:

fixes $f g :: 'a \Rightarrow \text{real}$
assumes $\text{integrable } M f$
assumes $\text{integrable } M g$
assumes $\text{indep-var borel } f \text{ borel } g$
shows $\text{covariance } f g = 0$
<proof>

lemma bienaymes-identity-2:

fixes $f :: 'b \Rightarrow 'a \Rightarrow \text{real}$
assumes $\text{finite } I$
assumes $\bigwedge i. i \in I \Rightarrow f i \in \text{borel-measurable } M$
assumes $\bigwedge i. i \in I \Rightarrow \text{integrable } M (\lambda \omega. f i \omega^2)$
shows $\text{variance } (\lambda \omega. (\sum i \in I. f i \omega)) =$
 $(\sum i \in I. \text{variance } (f i)) + (\sum i \in I. \sum j \in I - \{i\}. \text{covariance } (f i) (f j))$

$(f\ i)\ (f\ j)$
 $\langle proof \rangle$

theorem *bienaymes-identity-pairwise-indep:*

fixes $f :: 'b \Rightarrow 'a \Rightarrow real$
assumes *finite I*
assumes $\bigwedge i. i \in I \Rightarrow f\ i \in borel\text{-measurable}\ M$
assumes $\bigwedge i. i \in I \Rightarrow integrable\ M\ (\lambda\ \omega. f\ i\ \omega^{\wedge}2)$
assumes $\bigwedge i\ j. i \in I \Rightarrow j \in I \Rightarrow i \neq j \Rightarrow indep\text{-var}\ borel\ (f\ i)$
borel (f j)
shows $variance\ (\lambda\ \omega. (\sum i \in I. f\ i\ \omega)) = (\sum i \in I. variance\ (f\ i))$
 $\langle proof \rangle$

lemma *indep-var-from-indep-vars:*

assumes $i \neq j$
assumes *indep-vars* $(\lambda\ -. M')\ f\ \{i, j\}$
shows *indep-var* $M'\ (f\ i)\ M'\ (f\ j)$
 $\langle proof \rangle$

lemma *bienaymes-identity-pairwise-indep-2:*

fixes $f :: 'b \Rightarrow 'a \Rightarrow real$
assumes *finite I*
assumes $\bigwedge i. i \in I \Rightarrow f\ i \in borel\text{-measurable}\ M$
assumes $\bigwedge i. i \in I \Rightarrow integrable\ M\ (\lambda\ \omega. f\ i\ \omega^{\wedge}2)$
assumes $\bigwedge J. J \subseteq I \Rightarrow card\ J = 2 \Rightarrow indep\text{-vars}\ (\lambda\ -. borel)\ f\ J$
shows $variance\ (\lambda\ \omega. (\sum i \in I. f\ i\ \omega)) = (\sum i \in I. variance\ (f\ i))$
 $\langle proof \rangle$

lemma *bienaymes-identity-full-indep:*

fixes $f :: 'b \Rightarrow 'a \Rightarrow real$
assumes *finite I*
assumes $\bigwedge i. i \in I \Rightarrow f\ i \in borel\text{-measurable}\ M$
assumes $\bigwedge i. i \in I \Rightarrow integrable\ M\ (\lambda\ \omega. f\ i\ \omega^{\wedge}2)$
assumes *indep-vars* $(\lambda\ -. borel)\ f\ I$
shows $variance\ (\lambda\ \omega. (\sum i \in I. f\ i\ \omega)) = (\sum i \in I. variance\ (f\ i))$
 $\langle proof \rangle$

end

end

4 Cantelli's Inequality

Cantelli's inequality [3] is an improvement of Chebyshev's inequality for one-sided tail bounds.

theory *Cantelli-Inequality*

imports *HOL-Probability.Probability*
begin

context *prob-space*

begin

lemma *cantelli-arith:*

assumes $a > (0::real)$

shows $(V + (V / a)^2) / (a + (V / a))^2 = V / (a^2 + V)$ (**is**
?L = ?R)

<proof>

theorem *cantelli-inequality:*

assumes [*measurable*]: *random-variable* *borel* Z

assumes *intZsq*: *integrable* $M (\lambda z. Z z^2)$

assumes $a > 0$

shows $\text{prob} \{z \in \text{space } M. Z z - \text{expectation } Z \geq a\} \leq$
variance } Z / (a^2 + \text{variance } Z)

<proof>

corollary *cantelli-inequality-neg:*

assumes [*measurable*]: *random-variable* *borel* Z

assumes *intZsq*: *integrable* $M (\lambda z. Z z^2)$

assumes $a > 0$

shows $\text{prob} \{z \in \text{space } M. Z z - \text{expectation } Z \leq -a\} \leq$
variance } Z / (a^2 + \text{variance } Z)

<proof>

end

end

5 Efron-Stein Inequality

In this section we verify the Efron-Stein inequality. The verified theorem is stated as Efron-Stein inequality for non-symmetric functions by Steele [8]. However most textbook refer to this version as “the Efron-Stein inequality”. The original result that was shown by Efron and Stein is a tail bound for the variance of a symmetric functions of i.i.d. random variables [4].

theory *Efron-Stein-Inequality*

imports *Concentration-Inequalities-Preliminary*

begin

theorem *efron-stein-inequality-distr:*

fixes $f :: - \Rightarrow real$

assumes *finite* I

assumes $\bigwedge i. i \in I \implies \text{prob-space } (M i)$

assumes *integrable* ($PiMIM$) ($\lambda x. f x \hat{=} 2$) **and** *f-meas*: $f \in \text{borel-measurable}$
 $(PiMIM)$
shows *prob-space.variance* ($PiMIM$) $f \leq$
 $(\sum i \in I. (\int x. (f (\lambda j. x (j, False)) - f (\lambda j. x (j, j=i))) \hat{=} 2 \partial PiM$
 $(I \times UNIV) (M \circ fst))) / 2$
(is ?L ≤ ?R)
 $\langle \text{proof} \rangle$

theorem (*in prob-space*) *efron-stein-inequality-classic*:

fixes $f :: - \Rightarrow \text{real}$
assumes *finite* I
assumes *indep-vars* ($M' \circ fst$) $X (I \times (UNIV :: \text{bool set}))$
assumes $f \in \text{borel-measurable} (PiMIM')$
assumes *integrable* $M (\lambda \omega. f (\lambda i \in I. X (i, False) \omega) \hat{=} 2)$
assumes $\bigwedge i. i \in I \implies \text{distr } M (M' i) (X (i, True)) = \text{distr } M (M'$
 $i) (X (i, False))$
shows *variance* ($\lambda \omega. f (\lambda i \in I. X (i, False) \omega) \leq$
 $(\sum j \in I. \text{expectation} (\lambda \omega. (f (\lambda i \in I. X (i, False) \omega) - f (\lambda i \in I. X$
 $(i, i=j) \omega) \hat{=} 2)) / 2$
(is ?L ≤ ?R)
 $\langle \text{proof} \rangle$

end

6 McDiarmid's inequality

In this section we verify McDiarmid's inequality [6, Lemma 1.2]. In the source and also further sources sometimes refer to the result as the "independent bounded differences" inequality.

theory *McDiarmid-Inequality*

imports *Concentration-Inequalities-Preliminary*
begin

lemma *Collect-restr-cong*:

assumes $A = B$
assumes $\bigwedge x. x \in A \implies P x = Q x$
shows $\{x \in A. P x\} = \{x \in B. Q x\}$
 $\langle \text{proof} \rangle$

lemma *ineq-chain*:

fixes $h :: \text{nat} \Rightarrow \text{real}$
assumes $\bigwedge i. i < n \implies h (i+1) \leq h i$
shows $h n \leq h 0$
 $\langle \text{proof} \rangle$

lemma *restrict-subset-eq*:

assumes $A \subseteq B$
assumes $\text{restrict } f B = \text{restrict } g B$

shows $\text{restrict } f \ A = \text{restrict } g \ A$
 ⟨proof⟩

Bochner Integral version of Hoeffding's Lemma using *interval-bounded-random-variable.Hoeffding*

lemma (in *prob-space*) *Hoeffdings-lemma-bochner*:
assumes $l > 0$ **and** $E0$: *expectation* $f = 0$
assumes *random-variable borel* f
assumes $AE \ x \ \text{in } M$. $f \ x \in \{a..b::real\}$
shows *expectation* $(\lambda x. \exp (l * f \ x)) \leq \exp (l^2 * (b - a)^2 / 8)$ (is ?L ≤ ?R)
 ⟨proof⟩

lemma (in *prob-space*) *Hoeffdings-lemma-bochner-2*:
assumes $l > 0$ **and** $E0$: *expectation* $f = 0$
assumes *random-variable borel* f
assumes $\bigwedge x \ y. \{x,y\} \subseteq \text{space } M \implies |f \ x - f \ y| \leq (c::real)$
shows *expectation* $(\lambda x. \exp (l * f \ x)) \leq \exp (l^2 * c^2 / 8)$ (is ?L ≤ ?R)
 ⟨proof⟩

lemma (in *prob-space*) *Hoeffdings-lemma-bochner-3*:
assumes *expectation* $f = 0$
assumes *random-variable borel* f
assumes $\bigwedge x \ y. \{x,y\} \subseteq \text{space } M \implies |f \ x - f \ y| \leq (c::real)$
shows *expectation* $(\lambda x. \exp (l * f \ x)) \leq \exp (l^2 * c^2 / 8)$ (is ?L ≤ ?R)
 ⟨proof⟩

Version of *product-sigma-finite.product-integral-singleton* without the condition that $M \ i$ has to be sigma finite for all i :

lemma *product-integral-singleton*:
fixes $f :: - \Rightarrow -::\{banach, \text{second-countable-topology}\}$
assumes *sigma-finite-measure* $(M \ i)$
assumes $f \in \text{borel-measurable } (M \ i)$
shows $(\int x. f \ (x \ i) \ \partial(PiM \ \{i\} \ M)) = (\int x. f \ x \ \partial(M \ i))$ (is ?L = ?R)
 ⟨proof⟩

Version of *product-sigma-finite.product-integral-fold* without the condition that $M \ i$ has to be sigma finite for all i :

lemma *product-integral-fold*:
fixes $f :: - \Rightarrow -::\{banach, \text{second-countable-topology}\}$
assumes $\bigwedge i. \ i \in I \cup J \implies \text{sigma-finite-measure } (M \ i)$
assumes $I \cap J = \{\}$
assumes *finite* I
assumes *finite* J
assumes *integrable* $(PiM \ (I \cup J) \ M) \ f$
shows $(\int x. f \ x \ \partial(PiM \ (I \cup J) \ M)) = (\int x. (\int y. f \ (\text{merge } I \ J(x,y)) \ \partial(PiM \ J \ M)) \ \partial(PiM \ I \ M))$ (is ?L = ?R)

and *integrable* (PiM I M) ($\lambda x. (\int y. f (\text{merge } I J(x,y)) \partial \text{PiM } J M)$) (**is** ?I)
and *AE* x in PiM I M. *integrable* (PiM J M) ($\lambda y. f (\text{merge } I J(x,y))$) (**is** ?T)
 <proof>

lemma *product-integral-insert*:
fixes $f :: - \Rightarrow - :: \{\text{banach, second-countable-topology}\}$
assumes $\bigwedge k. k \in \{i\} \cup J \implies \text{sigma-finite-measure } (M k)$
assumes $i \notin J$
assumes *finite* J
assumes *integrable* (PiM (insert i J) M) f
shows $(\int x. f x \partial \text{PiM } (\text{insert } i J) M) = (\int x. (\int y. f (y(i := x)) \partial \text{PiM } J M) \partial M i)$ (**is** ?L = ?R)
 <proof>

lemma *product-integral-insert-rev*:
fixes $f :: - \Rightarrow - :: \{\text{banach, second-countable-topology}\}$
assumes $\bigwedge k. k \in \{i\} \cup J \implies \text{sigma-finite-measure } (M k)$
assumes $i \notin J$
assumes *finite* J
assumes *integrable* (PiM (insert i J) M) f
shows $(\int x. f x \partial \text{PiM } (\text{insert } i J) M) = (\int y. (\int x. f (y(i := x)) \partial M i) \partial \text{PiM } J M)$ (**is** ?L = ?R)
 <proof>

lemma *merge-empty[simp]*:
 $\text{merge } \{\} I (y,x) = \text{restrict } x I$
 $\text{merge } I \{\} (y,x) = \text{restrict } y I$
 <proof>

lemma *merge-cong*:
assumes $\text{restrict } x1 I = \text{restrict } x2 I$
assumes $\text{restrict } y1 J = \text{restrict } y2 J$
shows $\text{merge } I J (x1,y1) = \text{merge } I J (x2,y2)$
 <proof>

lemma *restrict-merge*:
 $\text{restrict } (\text{merge } I J x) K = \text{merge } (I \cap K) (J \cap K) x$
 <proof>

lemma *map-prod-measurable*:
assumes $f \in M \rightarrow_M M'$
assumes $g \in N \rightarrow_M N'$
shows $\text{map-prod } f g \in M \otimes_M N \rightarrow_M M' \otimes_M N'$
 <proof>

lemma *mc-diarmid-inequality-aux*:
fixes $f :: (\text{nat} \Rightarrow 'a) \Rightarrow \text{real}$

```

fixes n :: nat
assumes  $\bigwedge i. i < n \implies \text{prob-space } (M\ i)$ 
assumes  $\bigwedge i\ x\ y. i < n \implies \{x,y\} \subseteq \text{space } (PiM\ \{..\ < n\}\ M) \implies$ 
 $(\forall j \in \{..\ < n\} - \{i\}. x\ j = y\ j) \implies |f\ x - f\ y| \leq c\ i$ 
assumes f-meas:  $f \in \text{borel-measurable } (PiM\ \{..\ < n\}\ M)$  and  $\varepsilon\text{-gt-0}$ :
 $\varepsilon > 0$ 
shows  $\mathcal{P}(\omega \text{ in } PiM\ \{..\ < n\}\ M. f\ \omega - (\int \xi. f\ \xi\ \partial PiM\ \{..\ < n\}\ M) \geq$ 
 $\varepsilon) \leq \exp(-2 * \varepsilon^2 / (\sum i < n. (c\ i)^2))$ 
(is ?L ≤ ?R)
<proof>

```

```

theorem mc-diarmid-inequality-distr:
fixes f :: ('i  $\Rightarrow$  'a)  $\Rightarrow$  real
assumes finite I
assumes  $\bigwedge i. i \in I \implies \text{prob-space } (M\ i)$ 
assumes  $\bigwedge i\ x\ y. i \in I \implies \{x,y\} \subseteq \text{space } (PiM\ I\ M) \implies (\forall j \in I - \{i\}.$ 
 $x\ j = y\ j) \implies |f\ x - f\ y| \leq c\ i$ 
assumes f-meas:  $f \in \text{borel-measurable } (PiM\ I\ M)$  and  $\varepsilon\text{-gt-0}$ :  $\varepsilon > 0$ 
shows  $\mathcal{P}(\omega \text{ in } PiM\ I\ M. f\ \omega - (\int \xi. f\ \xi\ \partial PiM\ I\ M) \geq \varepsilon) \leq \exp$ 
 $(-2 * \varepsilon^2 / (\sum i \in I. (c\ i)^2))$ 
(is ?L ≤ ?R)
<proof>

```

```

lemma (in prob-space) mc-diarmid-inequality-classic:
fixes f :: ('i  $\Rightarrow$  'a)  $\Rightarrow$  real
assumes finite I
assumes indep-vars N X I
assumes  $\bigwedge i\ x\ y. i \in I \implies \{x,y\} \subseteq \text{space } (PiM\ I\ N) \implies (\forall j \in I - \{i\}.$ 
 $x\ j = y\ j) \implies |f\ x - f\ y| \leq c\ i$ 
assumes f-meas:  $f \in \text{borel-measurable } (PiM\ I\ N)$  and  $\varepsilon\text{-gt-0}$ :  $\varepsilon > 0$ 
shows  $\mathcal{P}(\omega \text{ in } M. f\ (\lambda i \in I. X\ i\ \omega) - (\int \xi. f\ (\lambda i \in I. X\ i\ \xi)\ \partial M) \geq$ 
 $\varepsilon) \leq \exp(-2 * \varepsilon^2 / (\sum i \in I. (c\ i)^2))$ 
(is ?L ≤ ?R)
<proof>

```

end

7 Paley-Zygmund Inequality

This section proves slight improvements of the Paley-Zygmund Inequality [7]. Unfortunately, the improvements are on Wikipedia with no citation.

```

theory Paley-Zygmund-Inequality
imports Lp.Lp
begin

```

```

context prob-space
begin

```

theorem *paley-zygmund-inequality-holder:*
assumes $p: 1 < (p::real)$
assumes $rv: \text{random-variable borel } Z$
assumes $intZp: \text{integrable } M (\lambda z. |Z z| \text{ powr } p)$
assumes $t: \vartheta \leq 1$
assumes $ZAEPoS: AE z \text{ in } M. Z z \geq 0$
shows
 $(\text{expectation } (\lambda x. |Z x - \vartheta * \text{expectation } Z| \text{ powr } p) \text{ powr } (1 / (p-1))) * \text{prob } \{z \in \text{space } M. Z z > \vartheta * \text{expectation } Z\} \geq ((1-\vartheta) \text{ powr } (p / (p-1)) * \text{expectation } Z \text{ powr } (p / (p-1)))$
 $\langle \text{proof} \rangle$

corollary *paley-zygmund-inequality:*
assumes $rv: \text{random-variable borel } Z$
assumes $intZsq: \text{integrable } M (\lambda z. (Z z)^2)$
assumes $t: \vartheta \leq 1$
assumes $Zpos: \bigwedge z. z \in \text{space } M \implies Z z \geq 0$
shows
 $(\text{variance } Z + (1-\vartheta)^2 * (\text{expectation } Z)^2) * \text{prob } \{z \in \text{space } M. Z z > \vartheta * \text{expectation } Z\} \geq (1-\vartheta)^2 * (\text{expectation } Z)^2$
 $\langle \text{proof} \rangle$

end

end

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