STAT 830
Convergence of RVs

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STAT 830 — Fall 2011

Purposes of These Notes

- Distinguish convergence in distribution from other modes of convergence.
- Describe which modes of convergence imply which others.

Intuition pp 71-72

- Think about a sequence X_n and a possible limit X:
  - X_n converges in distribution to X depends *only* on marginal distributions of individual X_n and X.
  - Convergence in probability and pth mean depends only on sequence of bivariate joint distributions of (X_n, X).
  - Convergence almost surely depends on joint distribution of all the variables: $X_1, X_2, \ldots, X$.
- All depend on scaling!
- In an iid sequence $\bar{X}_n$ converges in all senses to $\mu = \mathrm{E}(X_1)$ (for pth mean add the hypothesis that $\mathrm{E}(|X_1|^p) < \infty$. Y.
- In addition $\sqrt{n}(\bar{X}_n - \mu)$ converges in distribution to a normal random variable if $\mathrm{Var}(X_1) < \infty$.
- But not in any of the other senses of convergence.

Relation between modes of convergence

- If X_n converges to X almost surely then X_n converges to X in probability.
- If X_n converges to X in probability then X_n converges to X in distribution.
- If X_n converges to X in pth mean for some $p > 0$ then X_n converges to X in probability.
- If X_n converges to X in probability and the sequence is *uniformly pth power integrable* then X_n converges to X in pth mean.
- **Def'n**: Uniformly pth power integrable means

$$\lim_{M\to\infty} \sup\{\mathrm{E}(|X_n|^p 1(|X_n| > M)) = 0.$$

- Most easily checked by: $\exists\delta > 0$ such that

$$\sup\{\mathrm{E}(|X_n|^{p+\delta}) < \infty.$$