**1.** (Durrett 2.5.3) Define  $Y_n = \frac{\sin(n\pi t)}{n} X_n$ , then since  $X_n \sim N(0,1)$  we get  $\operatorname{Var} Y_n = \frac{|\sin(n\pi t)|}{n^2} \leq 1/n^2$ . Then since  $\sum_1^{\infty} \operatorname{Var} Y_n < \infty$  and  $Y_1, Y_2, \ldots$  are independent, we get from the consequence of Kolmogorov maximal inequality that  $\sum_1^{\infty} Y_n$  converges almost surely.

2. (Durrett 2.5.6) We can use the Kolmogorov 3-series test. First notice,

$$E[\psi(X_n)] = E[|X_n|1_{|X_n|>1} + X_n^2 1_{|X_n|<1}]$$

So in particular by the comparison test,

$$\sum_{1}^{\infty} E\left[|X_{n}|1_{|X_{n}|>1}\right] < \infty \qquad \qquad \sum_{1}^{\infty} E\left[X_{n}^{2}1_{|X_{n}|\leq 1}\right] < \infty$$

The latter is the series  $\sum_{1}^{\infty} \text{Var}(X_n 1_{|X_n| < 1})$ , another of the series is relatively free, namely

$$\sum_{1}^{\infty} P(|X_n| > 1) \le \sum_{1} E\left[|X_n| 1_{|X_n| > 1}\right] < \infty$$

Finally, we have  $|E[X_n 1_{|X_n|>1}]| \le E[|X_n 1_{|X_n|>1}]$ , so that  $\sum_{1}^{\infty} E[X_n 1_{|X_n|>1}]$  converges, since

$$0 = \sum_{1}^{N} E\left[X_{n}\right] = \sum_{1}^{N} E\left[X_{n} 1_{|X_{n}| > 1} + X_{n} 1_{|X_{n} \le 1|}\right] = \sum_{1}^{N} E\left[X_{n} 1_{|X_{n}| > 1}\right] + \sum_{1}^{N} E\left[X_{n} 1_{|X_{n} \le 1|}\right]$$

we have

$$\sum_{1}^{\infty} E\left[X_n 1_{|X_n \le 1|}\right] = -\sum_{1}^{\infty} E\left[X_n 1_{|X_n| > 1}\right] \in \mathbb{R}$$

Thus  $X_n$  satisfy the hypothesis of the Kolmogorov 3-series test and converge almost surely.

**3.** (Durrett 2.5.8) Write  $Y = \log^+|X_1|$  and assume first that  $EY = \infty$ , then for any c > 0, we find

$$EY = \int P(Y > t)dt = \sum_{1}^{\infty} \int_{(n-1)c}^{nc} P(Y > t)dt \le \sum_{1}^{\infty} cP(Y > nc)$$

which in turn implies that  $\sum_{1}^{\infty} P(\log^{+}|X_{n}| > nc) = \sum_{1}^{\infty} P(Y > nc) = \infty$ , therefore by Borel-Cantelli (ii), we find that  $P(X_{n} > e^{nc} \text{ i.o.}) = 1$ , so for any  $c \neq 0$  we find that  $|X_{n}|c^{n} > 1$  i.o. almost-surely. This of course implies that  $\sum_{1}^{\infty} |X_{n}|c^{n}$  has radius of convergence zero almost surely.

Now, conversely suppose  $E \log^+|X_n| < \infty$ , and let 0 < c < 1, then choose  $\gamma > 0$  so that  $e^{\gamma}c < 1$  we get another layer cake estimate,

$$EY = \int P(Y > t)dt = \sum_{1}^{\infty} \int_{(n-1)\gamma}^{n\gamma} P(Y > t)dt \ge \sum_{1}^{\infty} \gamma P(Y > n\gamma)$$

Since  $\sum_{1}^{\infty} \gamma P(Y > n\gamma) = \sum_{1}^{\infty} P(\log^{+}|X_{n}| > n\gamma)$  by Borel Cantelli, we have  $|X_{n}| > e^{n\gamma}$  only finitely many times. Letting N so that  $n \geq N$  implies  $|X_{n}| \leq e^{n\gamma}$ , we get

$$\sum_{1}^{\infty} c^{n} |X_{n}| \stackrel{\text{a.s.}}{\leq} \sum_{1}^{N-1} c^{n} |X_{n}| + \sum_{N}^{\infty} (ce^{\gamma})^{n} < \infty$$

So that the series converges a.s. for any c < 1. Now letting c > 1, we once again use Borel-Cantelli (ii), we first check the hypothesis

$$\sum_{1}^{\infty} P(|X_n|c^n > 1) = \sum_{1}^{\infty} P(|X_1| > \frac{1}{c^n}) = \infty$$

Since  $\lim_{n\to\infty} P(|X_1| > \frac{1}{c^n}) = P(X_1 \neq 0) > 0$  by assumption, it follows that  $P(|X_n|c^n > 1 \text{ i.o.}) = 1$ , so the series diverges a.s. for c > 1.

**4.** (Durrett 3.2.9) We can use a mesh to show that  $||F - F_n||_u = 0$ . Namely, let  $\epsilon > 0$ , then we can choose  $x_0$  so that  $F(x_1) < \epsilon$  and y so that  $F(y) > 1 - \epsilon$ , then F is uniformly continuous on  $[x_0, y]$ , so in particular there is some  $\delta > 0$  such that  $|w - z| < \delta \implies |F(w) - F(z)| < \epsilon$  on  $[x_0, y]$ , then take  $x_j = x_0 + j\frac{\delta}{2}$  for  $j \in \{1, \ldots, N-1\}$ , so that  $y - \delta/2 < x_{N-1} \le y$  and denote  $y = x_N$ . Then take  $M_j$ , so that  $n \ge M_j \implies |F_n(x_j) - F(x_j)| < \epsilon$  and define  $M = \max_{0 \le j \le N} M_j$ . It follows that for  $n \ge N$ , we have for  $x \in (-\infty, x_0)$ ,

$$F_n(x) \le F_n(x_0) < F(x_0) + \epsilon < 2\epsilon$$

and for  $x \in (x_N, \infty)$ 

$$F_n(x) \ge F_n(x_N) > F(x_N) - \epsilon > 1 - 2\epsilon$$

and on these ranges  $F(x) < \epsilon$  and  $F(x) > 1 - \epsilon$  respectively, meaning  $|F(x) - F_n(x)| < 2\epsilon$ . Finally, suppose that  $x \in [x_{j-1}, x_j]$  for some j, then we have

$$F(x) - 2\epsilon < F(x_{j-1}) - \epsilon \le F_n(x_{j-1}) \le F_n(x) \le F_n(x_j) \le F(x_j) + \epsilon < F(x) + 2\epsilon$$

So that  $||F - F_n||_u < 2\epsilon$  and since  $\epsilon$  was arbitrary,  $||F - F_n||_u \to 0$ .