## Homework 1: Outline solutions

1. (a) Draw a picture: easiest to separate into  $x < y$ ,  $v < 0$  and  $x \ge y$ ,  $v \ge 0$ . The the range of values  $(v, w)$  of  $(V, W)$  is

$$
\{(v, w): -1 \le v < 0, -v \le w \le 1\} \cup \{(v, w): 0 \le v \le 1, v \le w \le 1\}
$$

(b) If transforming the density, again separate into  $x > y$  and  $x < y$ . The e.g. on  $x > y$  inverse transformation is  $X = W, Y = W - V$ , note this mapping is continuously differentable, and don't forget the Jacobian (even though it turns out to be 1).

Easier, is direct use of the distribution function

$$
P(V \le v, W \le w) = \Pr(X - Y \le v, \max(X, Y) \le w)
$$
  
=  $\Pr(Y \ge X - v, X \le w, Y \le W)$   
=  $\begin{cases} w^2 - \frac{1}{2}(w - v)^2, & \text{if } 0 \le v \le w \le 1 \\ \frac{1}{2}(v + w)^2, & \text{if } -1 \le v < 0, 0 < -v \le w \le 1. \end{cases}$ 

Use your picture!

Differentiating w.r.t v and w gives the uniform density 1 on the range of  $(V, W)$ .

(c) No, they cannot be independent since the ranges are not separable: we know  $|v| \leq w$ .

2. (a) The sequence of random variables  $V_n$  converges in distribution to  $U(0, 1)$  since

 $Pr(V_n \leq x) = |nx|/n \rightarrow x \text{ as } n \rightarrow \infty$ 

for  $0 < x < 1$ . (Draw a picture!)

For convergence in probability there is insufficient information to say. We do not know the joint dsn of the  $V_n$  nor even that they are defined on the same probability space.

(b) There are various ways to do this. This one comes from Ferguson.

Let  $\epsilon > 0$  and  $k > 2/\epsilon$  (an integer). Since F is continuous, can choose  $x_j$  s.t.  $F(x_j) = j/k$ ,  $j = 1, ..., k - 1$ . Since  $F_n(x_j) \to F(x_j)$ ,  $\exists N_j$  s.t.  $|F_n(x_j) - F(x_j)| < 1/k$  for all  $n > N_j$ . Now consider  $n > N = \max(N_1, ..., N_{k-1}),$  and  $x_j \le x < x_{j+1}$ , with  $x_k = \infty$ :

$$
F_n(x) \le F_n(x_{j+1}) \le F(x_{j+1}) + 1/k \le F(x) + 2/k
$$
  

$$
F_n(x) \ge F_n(x_j) \ge F(x_j) - 1/k \ge F(x) - 2/k
$$

Hence  $|F_n(x) - F(x)| \leq 2/k < \epsilon$  for all  $n > N$  and for all x.

3. (a) Checking the integral

$$
\int_0^1 \int_0^1 f_\theta(v, w) dv dw = 1 \text{ for all real } \theta
$$

we see that all we need is  $f_{\theta}(v, w) \ge 0$  for all  $(v, w)$  in unit square, which by monotonicity of  $(1 - 2v)$ will hold iff it holds at the corners  $(0,0), (0,1), (1,0)$  and  $(1,1)$  of the square, or for  $|\theta| < 1$ .

(b) The corresponding distribution function  $F_{\theta}$  is given by

$$
F_{\theta}(v, w) = \int_0^v \int_0^w f_{\theta}(x, y) dx dy
$$
  
=  $vw(1 + \theta(1 - v)(1 - w))$ 

Note that

$$
F_{\theta}(v,1) = v \quad \text{and} \quad F_{\theta}(1,w) = w,
$$

so each of V and W has a uniform  $U(0, 1)$  marginal distribution.

(c) Now  $E(V) = E(W) = 1/2$  and  $var(V) = var(W) = 1/12$ , Then direct integration gives

$$
Cov(V, W) = \int_0^1 \int_0^1 v w f_{\theta}(v, w) dv dw - (1/2)^2 = \theta/36
$$

and hence

$$
\rho(V, W) = \frac{\text{Cov}(V, W)}{\sqrt{\text{var}(V)\text{var}(W)}} = \theta/3
$$

Thus  $|\rho(V, W)| \leq 1/3$ , which limits the range of correlation between two  $U(0, 1)$  random variables that can be provided by this simple family.

4. (a)  $E(Y_n) = \theta/2$  and  $var(Y_n) = \theta^2/12$ , hence immendiately by the CLT,  $V_n$  does converge in distribution, and it converges to  $N(0, \theta^2/12)$ .

(b) 
$$
Pr(Y_n) > y) = 1 - y/\theta
$$
, so  $Pr(min(Y_n) > y) = (1 - y/\theta)^n$ . Then  

$$
P(W_n > w) = Pr(min(Y_n) > y/n) = (1 - y/n\theta)^n
$$

$$
\rightarrow \exp(-y/\theta) \text{ as } n \rightarrow \infty
$$

Thus  $W_n$  does converge in distribution. It converges to an exponential random variable with mean  $\theta$ .