12 bivariate random variables marginal  $\longrightarrow f_X(X) \cdot dX = P[X \in (X, X+dX)] = P[X \in (X, X+dX), Y \in R] = ( f(X,Y) \cdot dY) dX$  $\frac{distribution}{distribution} > f_{Y}(Y)dY = P[Y \in (Y, Y+dY)] = P[Y \in (Y, Y+dY), X \in R] = \int_{-\infty}^{\infty} f(x, Y)dx)dY$ conditional sfx/y(x/y) dx = P[XE(x,x+dx)/Y=y] = f(x,y) dx
ensity { From P(x,y) = P(y) · P(x/y) => P(x/y) = P(x,y)/P(y) }  $\frac{distribution - fy[x(y|x)dy = P[y \in (y,y+dy)]/x=x] = \frac{f(x,y)dx}{fx(x)}}{fx(x)}$ fx(x) · fx/x(x/x)=fx(x)·fx/x(x/x) Bayes P(x,y)=P(y)·P(x/y) >>> P(x)·P(y/x)=P(y)·P(x/y)
P(y/x)=P(x/y)·P(y/x) P(Y/x) = P(x/y) · P(y)/P(x) = f(x/y) · f(y)/F(x) = f(x/y) · f(y)/F(x) expectation and variance  $\int Mx = E[x] = \iint x \cdot f(x, y) dx dy = \int x \cdot f_x(x) dx$ ( ox = var[X] = E[(x-Mx)2] = S(x-Mx)2.f(x,y) dx dy = S(x-Mx)2.f(x) dx  $MY = E[Y] = MY \cdot f(x,y) dxdy = MY \cdot f(Y) dy$ (oy= var[Y]= E[(+-HY)2]= [(Y-HY)2f(x,Y)dxdy= [(Y-HY)2f(x)dy conditional expectation and conditional variance  $m_{X}(y) = E[x/y=y] = \int x \cdot f_{XN}(x/y) dx = \int x \cdot f(x,y) dx / f_{Y}(y)$  $\left(\sigma_{X}(y) = E_{[(x-m_{X}(y))^{2}]} = \int (x-m_{X}(y))^{2} f_{X|Y}(x,y) dx = \int (x-m_{X}(y))^{2} f_{(x,y)} dx / f_{Y}(y)$  $\{m_{Y(x)} = E[Y/X=x] = \} y \cdot f_{Y|X}(y|x) dy = \} y \cdot f(x,y) dy / f_{X(x)}$  $(\sigma_{Y}(x) = E[y-m_{Y}(x)/x=x] = \int (y-m_{Y}(x))^{2} f_{Y}(x(y)x) dy = \int (y-m_{Y}(x))^{2} f(x,y) dy/f_{X}(x)$ Law of Iterated Expectations E[X] = E[E[X|Y=Y]] and var[X] = E[var[X|Y=Y]] + var[E[X|Y=Y]] E[Y]=E[E[Y|X=x]] and var[Y]=E[var[Y|X=x]]+var[E[Y|X=x]]