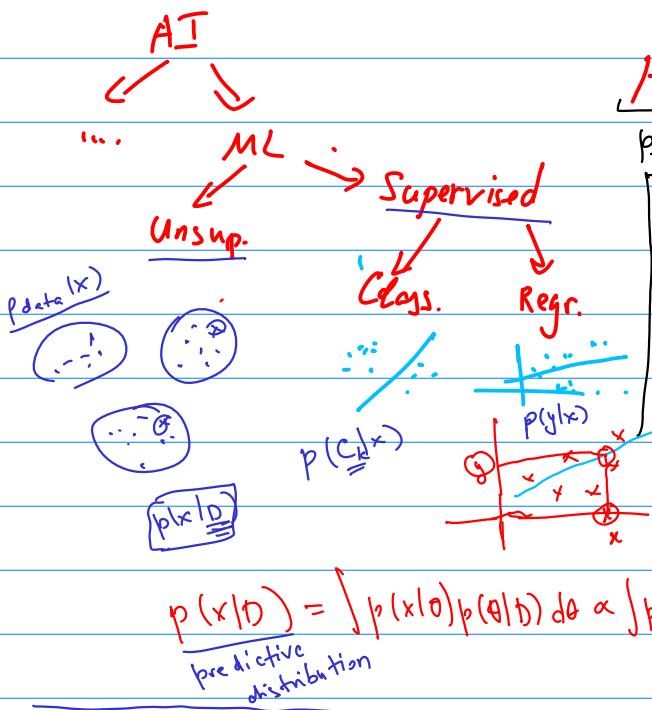


# Bienvenue à l'UData School 2018

Boîte à outils



$$p(\theta|D) = \frac{p(D|\theta)p(\theta)}{\int p(D|\theta)p(\theta)d\theta}$$

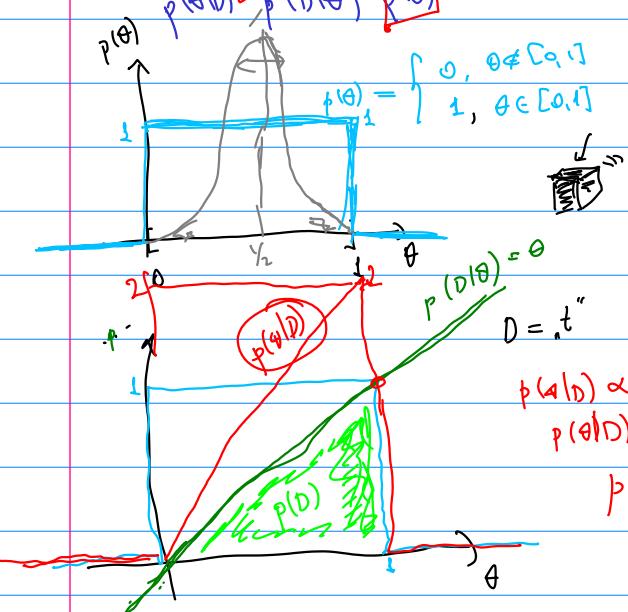
posterior

$$\text{KL: } p(D|\theta) \xrightarrow[\theta]{} \max \quad \theta_{\text{KL}}$$

$$\text{MAP: } p(\theta|D) \xrightarrow[\theta]{} \max \quad \theta_{\text{MAP}} \quad \propto \frac{p(\theta)}{p(D|\theta)}$$

$$\begin{aligned} p(x|D) &= \int p(x,\theta|D) d\theta = \\ &= \int \underbrace{p(\theta|D)}_{\text{posterior}} \cdot \underbrace{p(x|\theta)}_{\text{likelihood}} d\theta \end{aligned}$$

$\theta$ - "paramètres opéra"	$D = t$	Bernoulli trials
$p(b \theta)$	$p(t \theta) = \theta$	$p(d \theta) = \theta^n (1-\theta)^{n_h}$
$\theta_{\text{KL}} = \frac{n_t}{n_h + n_t}$	$p(t t,t t \theta) = \theta^{n_t} (1-\theta)^{n_h}$	
$\theta_{\text{KL}} = \arg \max_\theta p(D \theta)$	$n_t \cdot \theta^{n_t-1} (1-\theta)^{n_h} - n_h \cdot (1-\theta)^{n_h-1} \theta^{n_t} = 0$	
$p(b \theta) = \frac{\theta^2}{2} \rightarrow \max$	$\theta^{n_t-1} (1-\theta)^{n_h-1} (n_t \cdot (1-\theta) - n_h \cdot \theta) = 0$	
$D = \underline{t,t}: \quad \theta_{\text{KL}} = 1$	$\theta = 0 \quad \theta = 1$	$\theta_{\text{MAP}} = \frac{n_t}{n_h + n_t}$
$p(\theta b) \propto p(D \theta) \cdot p(\theta)$		



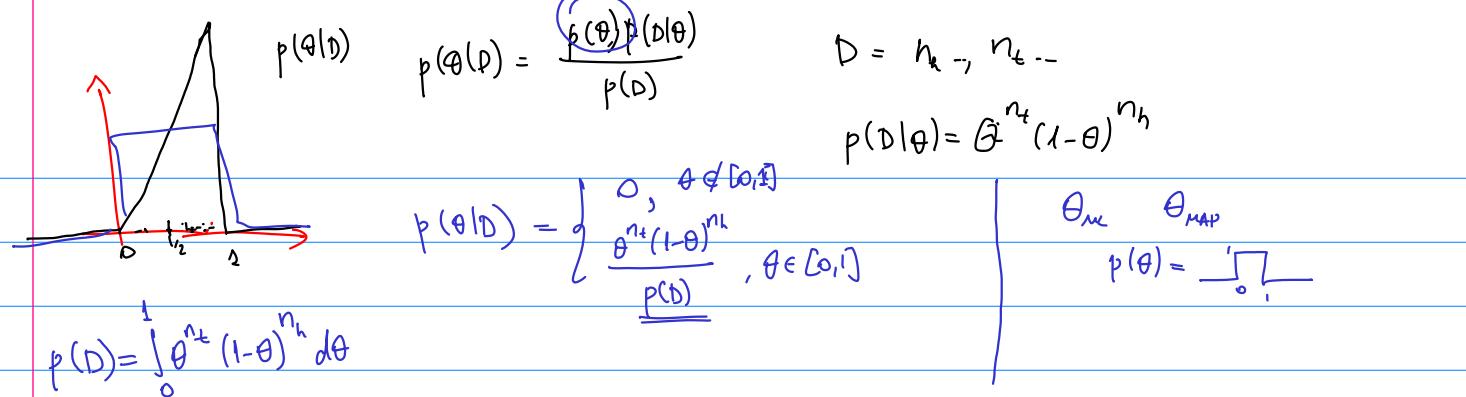
$$\begin{aligned} p(D|\theta) &= \theta^{n_t} (1-\theta)^{n_h} && \text{- likelihood} \\ p(\theta|D) &\propto \begin{cases} \theta^{n_t} (1-\theta)^{n_h}, & \theta \in [0,1] \\ 0, & \theta \notin [0,1] \end{cases} \end{aligned}$$

$$\theta_{\text{MAP}} = \arg \max_\theta p(\theta|D) = \frac{n_t}{n_h + n_t}$$

$$p(\theta|D) = \frac{p(\theta)p(D|\theta)}{\int p(\theta)p(D|\theta)d\theta}$$

$$p(D) = \int p(\theta)p(\theta|D)d\theta$$





$$B(\alpha, \beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)} = \frac{(\alpha-1)!(\beta-1)!}{(\alpha+\beta-1)!}$$

$$\Gamma(n) = (n-1)!$$

$$p(D) = \int_0^1 \theta^{n_t}(1-\theta)^{n_h} d\theta = \frac{n_t! n_h!}{(n_t+n_h+1)!}$$

$$p(t|D) = \int p(t, \theta|D) d\theta = \int p(t|\theta, D) p(\theta|D) d\theta =$$

$$= \int_0^1 \theta \cdot \frac{(n_t+n_h+1)!}{n_t! n_h!} \cdot \theta^{n_t} (1-\theta)^{n_h} d\theta = \dots \int_0^1 \theta^{n_t+1} (1-\theta)^{n_h} d\theta = \frac{(n_t+n_h+1)!}{n_t! n_h!} \cdot \frac{(n_t+1)! n_h!}{(n_t+n_h+2)!} = \frac{n_t+1}{n_t+n_h+2}$$

probabilistic language model

$$\Theta_{MAP} \rightarrow p(t|D) \quad \text{Bayesian smoothing}$$

$$\text{K gelen } \bar{\theta} = (\theta_1, \theta_2, \dots, \theta_k) \quad (\sum \theta_i = 1, \theta_i > 0)$$

$$p(D|\bar{\theta}) = \theta_1^{n_1} \theta_2^{n_2} \dots \theta_k^{n_k} \quad p(\bar{\theta}) = 1, \bar{\theta} \in \dots$$

$$p(i|t) = \frac{n_{i+1}}{\sum n_j + k}$$

$$p(w | \dots)$$

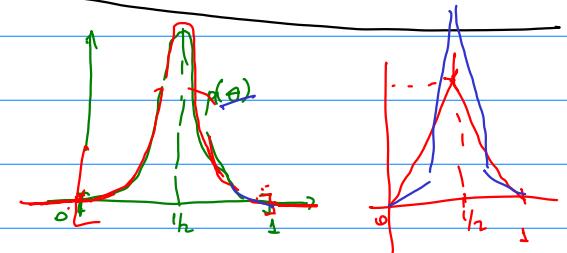
$$\begin{aligned} \text{n-grams} & \# \{ \text{beam\_output} \} \\ & \# \{ \text{redundant words} \} \end{aligned}$$

$$p([w_1 w_2 \dots] | \bar{\theta}) = 0$$

$$p(D|\theta) = \theta^{n_t} (1-\theta)^{n_h}$$

$$p(\theta|D), \Theta_{MAP}, p(t|D)$$

$$p(\theta) \times p(D|\theta) \propto p(\theta|D)$$



$$p(\theta|D) \cdot p(D|\theta) \propto p(\theta|D) = p(\theta|\bar{\theta}) \quad \therefore p(D|\theta) \propto p(\theta|\bar{\theta})$$

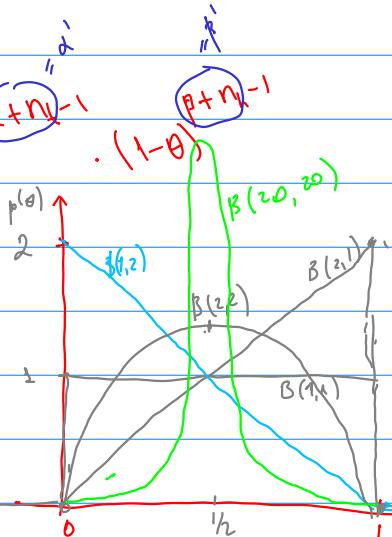
Conjugate priors: comenzando  $p(\theta|d)$  relativamente simple. Tanto  $p(D|\theta)$ ,  
 esas  $p(\theta|d) \cdot p(D|\theta) \propto p(\theta|d')$

$p(D|\theta)$

$$\frac{p(\theta) \cdot \theta^{n_1} \cdot (1-\theta)^{n_2}}{\theta^{2-1} \cdot (1-\theta)^{1-1} \cdot \theta^{n_1} \cdot (1-\theta)^{n_2}} \approx \frac{p(\theta)}{\theta^{2-1} \cdot (1-\theta)^{1-1}}$$

beta distribution

$$p(\theta) = \begin{cases} \frac{\theta^{\alpha-1} (1-\theta)^{\beta-1}}{B(\alpha, \beta)}, & \theta \in [0, 1] \\ 0, & \theta \notin [0, 1] \end{cases}$$



$\delta(1,1) \rightarrow D = \{n_1, n_2\} \rightarrow p(\theta|D) = B(n_1+1, n_2+1)$   
 $p(D|\theta) = \theta^{n_1} (1-\theta)^{n_2}$

$$p(\theta) = B(20, 20)$$

$$\delta(x; \frac{1}{2}, \frac{1}{2}) \propto x^{\frac{1}{2}} (1-x)^{\frac{1}{2}} = \frac{1}{\sqrt{x(1-x)}}$$

$$(1-\theta)^{\alpha-1} \rightarrow \max_{\theta}$$

$$\theta^* = \frac{\alpha-1}{\alpha + \beta - 2}$$

$$p(D|\bar{\theta}) = \theta_1^{n_1} \theta_2^{n_2} \dots \theta_k^{n_k}$$

$$p(\bar{\theta} | \bar{\theta}) = p(D|\bar{\theta}) \propto p(\bar{\theta}; \bar{\theta} + \bar{n})$$

$$p(\bar{\theta}) \propto \theta_1^{\alpha_1-1} \theta_2^{\alpha_2-1} \dots \theta_k^{\alpha_k-1}$$

$$\text{Dirichlet } p(\bar{\theta}) = \frac{\text{---}}{\text{Dir}(\bar{\theta})}$$

$$\text{Dir}(\alpha_1, \dots, \alpha_k) = \frac{\Gamma(\alpha_1) \cdot \Gamma(\alpha_2) \dots \Gamma(\alpha_k)}{\Gamma(\alpha_1 + \dots + \alpha_k)}$$

