

## Short handbook of the most useful formulas

### Definitions

$X_i(t)$ : offered traffic in class  $i$  in a scheduler

$Y_i^*(t)$  or  $\widehat{Y}_i(t)$  : available service for class  $i$  in a scheduler

$E()$ : average value operator

$V()$ : variance operator

### Alpha function

$$\alpha(t) = -\frac{E(X_i) - E(\widehat{Y}_i(t+d))}{\sqrt{V(X_i(t)) + V(\widehat{Y}_i(t+d))}}$$

### Delay tail

Given the absolute minimum of  $\alpha(t)$ , referred to as  $\alpha_{min}$ :

$$Pr(D \geq d) = e^{-\frac{\alpha_{min}^2 d}{2}}$$

### FIFO scheduler

$$\widehat{Y}_i(t) = Ct$$

$$E(\widehat{Y}_i(t)) = Ct$$

$$V(\widehat{Y}_i(t)) = 0$$

### Stric priority scheduler

$$\widehat{Y}_i(t) = Ct - \sum_{j=1}^{i-1} X_j(t)$$

$$E(\widehat{Y}_i(t)) = Ct - \sum_{j=1}^{i-1} E(X_j(t))$$

$$V(\widehat{Y}_i(t)) = \sum_{j=1}^{i-1} V(X_j(t))$$

### GPS scheduler

$$\widehat{Y}_i(t) = w_i Ct + \sum_{j \neq i} \frac{w_j}{1-w_j} (w_j Ct - X_j(t))$$

$$E(\widehat{Y}_i(t)) = w_i Ct + \sum_{j \neq i} \frac{w_j}{1-w_j} E(w_j Ct - X_j(t))$$

$$V(\widehat{Y}_i(t)) = \sum_{j \neq i} \left( \frac{w_j}{1-w_j} \right)^2 V(X_j(t))$$

### EDF scheduler

$$\widehat{Y}_i(t) = Ct - \sum_{j \neq i} X_j(t - \max(0, \delta_j - \delta_i))$$

$$E(\widehat{Y}_i(t)) = Ct - \sum_{j \neq i} E(X_j(t - \max(0, \delta_j - \delta_i)))$$

$$V(\widehat{Y}_i(t)) = \sum_{j \neq i} V(X_j(t - \max(0, \delta_j - \delta_i)))$$

### FIFO scheduler with SRD traffic

A FIFO scheduler with capacity  $C$ , offered  $N$  SRD traffic flows, where each traffic flow has parameters  $r$ ,  $b$ , and  $H=1/2$ . The requested SLA is  $(d,p)$ .

Delay tail

$$Pr(D>d) = \exp\left(-2 \frac{C(C-Nr)}{Nrb} d\right)$$

Average delay

$$E(D) = \frac{Nrb}{2C(C-Nr)}$$

Admission control

$$N \leq \frac{2C^2 d}{2Cdr - rb \ln(p)}$$

Resource provisioning

$$C \geq \frac{Nr}{2} + \sqrt{\left(\frac{Nr}{2}\right)^2 - \frac{Nrb \ln(p)}{2d}}$$

### FIFO scheduler with LRD traffic

A FIFO scheduler with capacity  $C$ , offered  $N$  SRD traffic flows, where each traffic flow has parameters  $r$ ,  $b$ , and  $H$ . The requested SLA is  $(d,p)$ .

Delay tail

$$Pr(D>d) = \exp\left(\frac{-1}{2Nrb} \frac{1}{H^{2H}(1-H)^{2-2H}} C^{2-2H} (C-Nr)^{2H} d^{2-2H}\right)$$

### Strict Priority (SP) scheduler with SRD traffic

A SP scheduler with capacity  $C$ , offered  $N_i$  SRD traffic flows in class  $i$ , where each traffic flow in class  $i$  has parameters  $r_i$ ,  $b_i$ , and  $H_i=1/2$ . The requested SLA for class  $i$  is  $(d_i,p_i)$ .

Delay tail

Define

$$A_i = \sum_{j=1}^i N_j r_j$$

(note that  $A_0=0$  )

$$B_i = \sum_{j=1}^i N_j r_j b_j$$

(note that  $B_0=0$  )

$$Pr(D_i>d_i) = \exp\left(-2 \frac{C-A_i}{B_i^2} [(C-A_{i-1})B_i - (C-A_i)B_{i-1}] d\right)$$

### Modification of the variance of a traffic flow when crossing a scheduler

$$V(X_{i,out}(t)) = \max(V(X_i(t)), \hat{Y}_i(t))$$

### Probability density of delay from delay tail

$$f_D(t) = \frac{-d}{dt} Pr(D > t)$$

### Probability density of end-to-end delay

When a traffic flow crosses a series of schedulers, the delay in scheduler  $i$  is  $D_i$ , and the total delay will be

$$D_{tot} = D_1 + D_2 + \dots + D_n,$$

where  $n$  is the number of schedulers. The probability density of  $D_{tot}$  is

$$f_{D_{tot}}(t) = f_{D_1}(t) \otimes f_{D_2}(t) \otimes \dots \otimes f_{D_n}(t)$$

where  $\otimes$  denotes the convolution operation.

If  $D_1, \dots, D_n$ , are identical exponential variables, that is

$$\forall i: f_{D_i}(t) = k e^{-kt}$$

then the probability density of the end-to-end delay is an Erlang

$$f_{D_{tot}}(t) = \frac{k^n t^{n-1}}{(n-1)!} e^{-kt}$$

If  $D_1, \dots, D_n$ , are exponential variables with different parameter  $k$ , that is

$$\forall i, j \neq i: f_{D_i}(t) = k_i e^{-k_i t}; f_{D_j}(t) = k_j e^{-k_j t}, k_i \neq k_j$$

then the probability density of end-to-end delay is an hyperexponential. In the case of  $n=2$ :

$$f_{D_{tot}}(t) = \frac{k_1 k_2}{k_2 - k_1} (e^{-k_1 t} - e^{-k_2 t})$$

### SRD model of VoIP traffic, codec with Voice Activity Detection

$$E(X(t)) = \frac{\lambda}{\lambda + \mu} P t$$

$$V(X(t)) = 2 \frac{\lambda \mu}{(\lambda + \mu)^3} P^2 t$$

where

$$\lambda = 1.538 \text{ s}^{-1}$$

$$\mu = 2.8571 \text{ s}^{-1}$$