

A Gaussian Approximation of Runtime Estimation in a Desktop Grid Project

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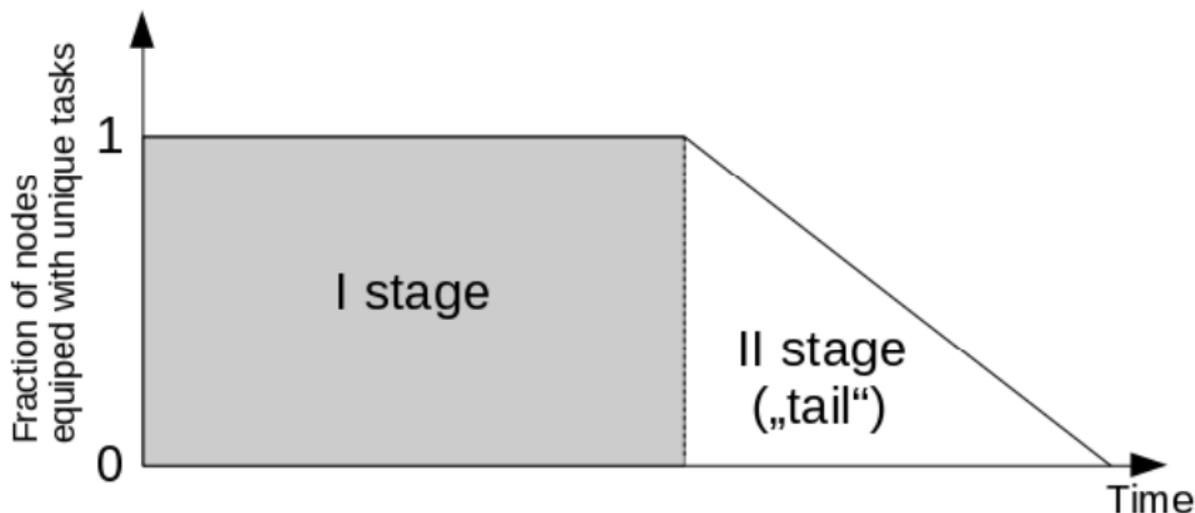
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- We propose a novel stochastic model describing the dynamics of a Desktop Grid project with many hosts and many workunits to be preformed.
- The model is based on the approximation of the basic process by means of the asymptotics of the superposed on-off sources both with light- and heavy-tailed distributions of the hosts' working sessions.
- It is well-known that, after an appropriate scaling, the limiting process describing the summary workload in the system turns out to be *Brownian motion* (BM), when the sources are light-tailed, while it becomes *Fractional Brownian motion* (FBM) when sources are heavy-tailed.
- Then the problem runtime estimation reduces to calculation of the hitting time of the given threshold the summary work to be done.

Two stages of project completion in a Desktop Grid:



- At the first stage the number of workunits is greater than the number of hosts, and thus each host can obtain at least one workunit.
- At the second stage, all available workunits are dispatched and there are available (idle) hosts.

On-Off model:



- Each host is characterized by the service rate R (the amount of processed work per unit time) and the following *on-off* process $\{I(t), t \geq 0\}$,

$$I(t) = \begin{cases} R, & t \in \text{on-period,} \\ 0, & t \in \text{off-period.} \end{cases} \quad (1)$$

- During an *on-period* each host is processing the work, while it keeps silence (inactive) during the following *off-period*. It is assumed that, for each host, on periods and off periods separately are iid.

- Assume that there are n types of hosts, and N_i is the number of the i th type of hosts.
- Each type $i = 1, \dots, n$ is characterized by the service rate R_i and corresponding *on-off* process $\{I^{(i)}(t), t \geq 0\}$,

$$I^{(i)}(t) = \begin{cases} R_i, & t \in \text{on-period,} \\ 0, & t \in \text{off-period.} \end{cases} \quad (2)$$

- The on-off processes generated by different hosts are assumed to be independent.

The cumulative processed work:

- The aggregated amount of the work (*cumulative processed work*) served by all hosts during time interval $[0, t]$ is defined as

$$A(t) = \int_0^t \left(\sum_{i=1}^n \sum_{k=1}^{N_i} I_k^{(i)}(u) \right) du,$$

where $I_k^{(i)}$ are independent copies of $I^{(i)}$, $i = 1, \dots, n$.

Distribution of on-off periods:

- The statistical behavior of the cumulative processed work crucially depends on the distribution of on-off periods. Let, for the i th type of hosts, F_{on}^i, F_{off}^i be the distribution functions of on- and off-period respectively. Denote by

$$\mu_{on}^i, \sigma_{on}^i, \mu_{off}^i, \sigma_{off}^i$$

the mean length and standard deviation of the i th type of on-, off-period, respectively, $i = 1, \dots, n$.

- We assume that the following conditions hold as $x \rightarrow \infty$:

$$\begin{aligned} 1 - F_{on}^i(x) &\sim x^{-\alpha_{on}^i} L_{on}^i(x), \quad \alpha_{on}^i \in (1, 2) \\ 1 - F_{off}^i(x) &\sim x^{-\alpha_{off}^i} L_{off}^i(x), \quad \alpha_{off}^i \in (1, 2) \end{aligned} \quad (3)$$

Gaussian approximation: theoretical motivation

Functional limit theorem argumentation leads to the following approximation for sufficiently large values N_i and T :

$$A(tT) \approx T \left(\sum_{i=1}^n R_i N_i \frac{\mu_{on}^i}{\mu_{on}^i + \mu_{off}^i} \right) t + \sum_{i=1}^n T^{H_i} R_i \sqrt{L_i(T) N_i c_i} B_{H_i}(t), \quad (4)$$

where c_i are positive constants, L_i are slowly varying at infinity functions (expressed in the terms of given parameters) and B_{H_i} are independent fractional Brownian motions with the Hurst parameters H_i defined as

$$H_i = \frac{3 - \min(\alpha_{on}^i, \alpha_{off}^i)}{2} \in \left(\frac{1}{2}, 1 \right), \quad i = 1, \dots, n.$$

- If $\sigma_{on}^i, \sigma_{off}^i < \infty$ for all $i = 1, \dots, n$, then the limiting process of the cumulative processed work becomes

$$T \left(\sum_{i=1}^n \frac{R_i N_i \mu_{on}^i}{\mu_{on}^i + \mu_{off}^i} \right) t + \left(\sqrt{T} \sum_{i=1}^n R_i \sqrt{N_i} c_i \right) W(t), \quad (5)$$

where $W(t)$ is a Wiener process, and constants c_i are expressed by the original predefined parameters

- Given above asymptotic results provide theoretical motivation to consider the following model:

$$A(t) = mt + \sigma X(t), \quad (6)$$

where m and σ are positive constants (can be estimated from observations); X is a centered Gaussian process, which describes random fluctuations of the cumulative processed work around the linearly increasing mean.

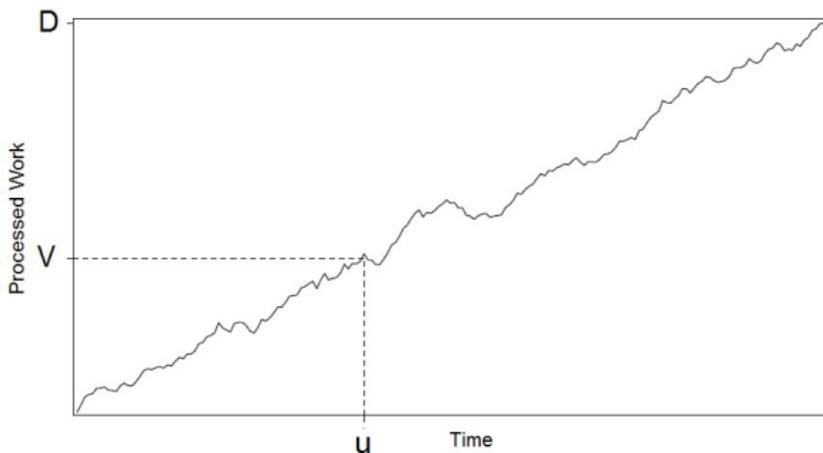
- The second term in expression (6) can be negative, so the trajectory of the process $A(t)$ in general is not monotonically increasing, what does not agree with the physical sense of the *processed work*.

- Let D be the required amount of work to be done in the Desktop Grid project, and we denote by τ_D the corresponding runtime of the project.
- The runtime can be defined as the corresponding *hitting time* of the process $A(t)$, that is,

$$\tau_D = \min\{t : A(t) \geq D\}. \quad (7)$$

- In case of Wiener process the probability density function of r. v. τ_D is available in explicit form.
- In case of fractional Brownian motion we are forced to use Monte-Carlo simulation to estimate expected value or other quantitative characteristics.

Runtime as a hitting time of Gaussian process



- From a practical point of view, we are interested in estimation the following conditional expectation

$$\gamma_{D,V}(u) = \mathbb{E}[\tau_D | \tau_V = u], \quad (8)$$

i. e., the expected runtime given that the fraction $V \leq D$ had been completed up to time instant u .

Concluding remarks:

- In this work a novel Gaussian model to describe the activity of the DG with a large number of hosts is suggested.
- Possible future research directions:
 - Validation of this model using statistical data describing real DG projects.
 - Comparison with some well-known direct statistical prediction techniques (for example based on autoregressive models).
 - Asymptotic analysis of the hitting time of fractional Brownian motion.

Thank you for attention!