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# Remarks on stochastic aspects on graphs

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#### 1. Introduction

For the description and solution of many problems in the area of operations research for which graph models have turned out to be an appropriate tool if problem data are fixed and known to enlarge the possibilities of application it is desirable to find corresponding formulations if variation of data has to be taken into consideration. With regard to this view-point problems including stochastic variation of data are of special interest and some efforts in combining stochastic and graphtheoretical aspects are cited in the references omitting, however, the large number of contributions using tools from reliability theory and stochastic programming (except [3] [5]). Of course, maximum flow and shortest path network formulations belong to the problems for which stochastic generalizations were considered (see e.g. [6], [9], [25] ), however, stochastic versions of project planning models seem to be mostly dealt with in this context. Defining a project to be a finite set of activities  $A = \{a_1, \dots, a_n\} \neq \emptyset$ with an irreflexive, asymmetric, transitive ordering relation  $\sigma$   $\subset$  AxA (which allows representation as an adjacency-relation on the arc-set of a graph at least after introducing dummy activities, see e.g [1], [20]) and nonnegative activity durations the problem is to determine the project duration which is yielded by maximizing over the sums of the durations of those activities which form maximal chains (with respect to  ${\mathfrak G}$  ). Assuming stochastic activity durations an intuitive idea is to replace the stochastic variables by their expected values and solve the resulting deterministic problem (using CPM), thus providing a rough estimate for the project duration mean. For this PERT-approach [18] several improvements have been suggested (see [4],[5],[10],[16],[22] ). This paper gives a description of a class of estimation possibilities containing some estimators known from the literature as special cases.

To get information about the distribution of the project duration one can also use simulation techniques (see e.g. [2], [12], [17], [23]) or try to reduce or decompose the project structure (see e.g. [2], [3] [4], [19]). Some decompositions of the project and properties of sub-

project systems are also given in this paper. A further generalization considering additional stochastic aspects influencing the project structure is not handled in this context (see e.g. [7], [12], [21] for an introduction to those problems).

# 2. Project-graphs

Let

(2.1) 
$$\widetilde{N}_{om} = (P, K, \varphi, X)$$

denote a (stochastic) project-graph (with deterministic structure) which means that  $(P,K,\phi)$  is a finite, directed, simple, acyclic, (weak) connected graph where P describes the set of nodes, K the set of arcs and  $\varphi = (\varphi^1, \varphi^2)$  with  $\varphi^i : \mathbb{K} \longrightarrow \mathbb{P}$ , i=1,2, the injective incidence mapping of the graph where  $\psi^1(k)$  resp.  $\psi^2(k)$  gives the starting-resp. end-point of k ∈ K. Additionally, (P,K, y) posesses two nodes s,tsP , called source, sink with  $\{k \in K | \varphi^2(k) = s\} = \emptyset$ ,  $\{k \in K | \varphi^1(k) = t\} = \emptyset$ . For such graphs a bijective mapping (called topological order of the nodes)  $G: P \longrightarrow \{0,1,\ldots,m\}, m := |P|-1$ , satisfying  $k \in K \implies G(\psi^{1}(k)) < G(\psi^{2}(k))$  (with G(s) = 0, G(t) = m) exists, and for ease of description, it is assumed that P := {0,1,...,m} is topologically ordered in the mentioned way. For  $\widetilde{K} \subset K$  let  $X_{\widetilde{K}}$  denote the vector of random variables  $X_{t}$ ,  $k \in \widetilde{K}$ defined on a probability space  $(\Omega, \mathcal{G}, \Pr)$  (sometimes the subscript  $\widetilde{K}$  is omitted) describing the arc passage times.  $X_k \in \mathcal{L}_1$ ,  $k \in K$  , and, if SK is an appropriate partition of K, independence  $X_{K_1}, \ldots, X_{K_p}$  , and the knowledge of the joint distribution of  $X_{K_i}$  ,  $i \in \{1, ..., \zeta\}$  is assumed (For ease of description often the special case  $|K_i| = 1$  is choosen. ). For p∈P one defines

(2.2)  $K(p) := \{k \in K | \psi^2(k) = p\}, B(p) := \{j \in P | \psi^1(k) = j, k \in K(p)\}$ 

<sup>1)</sup>  $\stackrel{\smile}{N}_{om}$  can be considered as graphtheoretical representation of a project planning model where no stochastic aspects influence the project structure.

from which one obtains the following subgraph of  $\stackrel{\sim}{N}_{om}$ 

(2.3) 
$$G_p = (P_p, K_p, \varphi_p)$$

with  $P_p:=\{0,1,\dots,p\}$ ,  $K_p:=\bigcup_{\mu=0}^p K(\mu)$  (as a partition of  $\bar{\phi}^1(P_p x P_p \wedge \phi(K))$ ),  $\phi_p=\phi/K_p$ . Of special interest are subproject-graphs for  $i,j\in P$ ,  $i\leq j$ , which (if existent) are denoted by

$$(2.4) Nij = (Pij, Kij, \varphiij, XKij)$$

where  $P_{i,j} \subset \{i,i+1,\ldots,j-1,j\}$   $(i,j\in P_{i,j})$  is source, sink of  $N_{i,j}$ ,  $K_{i,j} \subset \overline{\psi}^{-1}(P_{i,j} \times P_{i,j}) \cap \psi(K))$  have appropriately to be choosen to fit the project-graph conditions,  $\psi_{i,j} = \psi/K_{i,j}$ .  $\widetilde{N}_{op} = (\widetilde{P}_{op}, \widetilde{K}_{op}, \widetilde{\phi}_{op}, X_{\widetilde{K}_{op}})$  denotes the subproject-graph which is maximal in  $G_p$ . If  $W_{i,j}$  describes a route from i to j,  $i,j\in P$   $(W_{i,j})$  is a special subproject-graph),  $K(W_{i,j})$  the arc set of  $W_{i,j}$  and  $\mathcal{M}(N_{i,j})$  the set of routes  $W_{i,j}$  belonging to  $N_{i,j}$ , one can define

(2.5) 
$$L(N_{ij}) := \max_{Q_{r} f(N_{ij})} \sum_{k \in K(W_{ij})} \operatorname{proj}_{k}$$

which gives the  $N_{ij}$ -project passage time. Of special interest are

(2.6) 
$$L_p := L(\widetilde{N}_{OP})$$
 ,  $p \in P$ 

(with 
$$L_o := O$$
 ,  $L_m := L(\widetilde{N}_{om})$ ).

## 3. Subproject-graph systems

For  $p \in P$  ,  $p \ge 1$  , consider systems of subproject-graphs of the form

$$(3.1) \hspace{1cm} \mathfrak{N}_{\mathbf{p}} := \left\{ N_{\mathbf{ip}} \middle| N_{\mathbf{ip}}, \mathbf{icp}, \text{subproject-graph} \right\} \hspace{1cm} \text{with} \hspace{1cm} B(\mathfrak{N}_{\mathbf{p}}) := \left\{ \mathbf{iep} \middle| N_{\mathbf{ip}} \in \mathfrak{N}_{\mathbf{p}} \right\}$$

 $\eta$  is called proper subproject-graph system if  $^{2)}$ 

For graphs  $G_1 = (P_1, K_1)$ ,  $G_2 = (P_2, K_2)$  omitting the incidence mappings  $Y_1, Y_2$  (and stochastic vectors  $X_{K_1}, X_{K_2}$ ) the following notation is used  $G_1 \subset G_2$  iff  $P_1 \subset P_2$  and  $K_1 \subset K_2$  and  $G_1 \cap G_2 = (P_1 \cap P_2, K_1 \cap K_2)$  where, if no ambiguity can arise, only the relevant set in brackets is written.

$$(3.2) N_{i_1p}^1, N_{i_2p}^2 \in \mathcal{M}_p \Rightarrow N_{i_1p}^1 \cap N_{i_2p}^2 = \begin{cases} \{i,p\}, i_1=i_2=i \\ \{p\}, otherwise \end{cases}$$

(3.3) 
$$\forall W_{op} \exists N_{ip} \in \mathcal{M}_p : W_{op} = W_{oi} \cup W_{ip} \text{ with } W_{ip} \in N_{ip}$$

There always exist proper (subproject-graph) systems , e.g.  $\mathbb{T}_p = \{\widetilde{\mathbb{N}}_{op}\}$  is always proper.

Among the properties for proper systems one has

## (3.4) Theorem:

Let  $\mathcal{M}_p = \{N_{ip}\}$ ,  $\mathcal{M}_p' = \{N_{jp}'\}$  be two proper systems with  $\bigvee_{\mathcal{M}_p} K_{ip} \supset \bigvee_{\mathcal{M}_p'} K_{jp}'$ . Then there exists a proper system  $\mathcal{M}_p'' = \{N_{qp}'\}$  with

i)  $\bigvee_{\mathcal{M}_p'} K_{qp}'' = \bigvee_{\mathcal{M}_p} K_{ip}$  and

ii)  $\bigvee_{\mathcal{M}_p'} K_{qp}'' = \bigvee_{\mathcal{M}_p} K_{ip}$  and

For a proof see [11]. let be

the abreviated notation for the situation described in (3.4). The next theorem shows why proper subproject-graph systems are useful.

# (3.6) Theorem:

If  $\mathcal{M}_{q} = \{N_{iq}\}$  is a proper system,  $q \in P \setminus \{0\}$ , one has  $L_{q} = \max_{B(\mathcal{M}_{q})} \{ L_{i} + L(N_{iq}) \}$ 

For a proof one has to consider that for  $\omega \in \Omega$  and  $i_1 \in B(\mathfrak{N}_q)$  there exist  $\widehat{W}_{\text{oi}_1} \in \mathcal{M}(\widetilde{N}_{\text{oi}_1})$ ,  $\widehat{W}_{i_1q} \in \mathcal{M}(N_{i_1q})$  which maximize (2.6) (for  $p=i_1$ ) and (2.5) (for  $i=i_1,j=q$ ) for the realization  $X(\omega)$ , but  $\widehat{W}_{\text{oi}_1} \cup \widehat{W}_{i_1q} \in \mathcal{M}(\widetilde{N}_{\text{oq}})$ , thus

 $L_{i_{1}}^{(X(\omega))} + L(N_{i_{1}q}^{(X(\omega))} \leq L_{q}^{(X(\omega))}$ 

On the other side if  $\widehat{W}_{oq} \in \widehat{\mathcal{O}_{v}}(\widetilde{N}_{oq})$  maximizes (2.6) (for p=q) for  $X(\omega)$  because of (3.3) there exists  $N_{i_2q} \in \mathcal{M}_q$  such that

$$\begin{split} \widehat{\mathbb{W}}_{\mathrm{oq}} &= \mathbb{W}_{\mathrm{oi}_{2}} \cup \mathbb{W}_{\mathrm{i}_{2}\mathrm{q}} \quad \text{with} \quad \mathbb{W}_{\mathrm{i}_{2}\mathrm{q}} \in \mathcal{M}(\mathbb{N}_{\mathrm{i}_{2}\mathrm{q}}) \quad \text{and, of course,} \\ \mathbb{W}_{\mathrm{oi}_{2}} &\in \mathcal{M}(\widehat{\mathbb{N}}_{\mathrm{oi}_{2}}) \quad \text{, thus} \\ \mathbb{L}_{\mathrm{q}}(\mathbb{X}(\omega)) &= \sum_{\mathrm{k} \in \mathbb{K}(\mathbb{W}_{\mathrm{oi}_{2}})} \mathrm{proj}_{\mathrm{k}}(\mathbb{X}(\omega)) + \sum_{\mathrm{k} \in \mathbb{K}(\mathbb{W}_{\mathrm{i}_{2}\mathrm{q}})} \mathrm{proj}_{\mathrm{k}}(\mathbb{X}(\omega)) \\ &\leq \mathbb{L}_{\mathrm{i}_{2}}(\mathbb{X}(\omega)) + \mathbb{L}(\mathbb{N}_{\mathrm{i}_{2}\mathrm{q}})(\mathbb{X}(\omega)) \; . \end{split}$$
 Now, maximizing with respect to  $\mathbb{B}(\overline{\mathbb{N}}_{\mathrm{q}})$  gives the theorem.

# 4. Project passage time estimation

For ease description  $X_k$ ,  $k \in K$ , are assumed independent for the present. For  $Z \subseteq K$  let  $E_Z$  denote the integration with respect to  $X_Z$ , E (without subscript) the expectation. Then, for  $p \in P$ ,  $p \ge 1$ , if  $S_j$  are known optimistic estimators for the  $\widetilde{N}_{0j}$ -project passage times which means

(4.1) 
$$S_{i} \leq E(L_{i})$$
,  $j \in \{0,1,...,p-1\}$ , with  $S_{o} := 0$ 

one can define

$$(4.2) \qquad \qquad S(\mathfrak{M}_{p}, \mathbb{Z}_{p}, \mathbb{L}_{p}) := \mathbb{E}_{\mathbb{Z}_{p}} \max_{B(\mathfrak{M}_{p})} \left\{ S_{i} + \mathbb{E}_{\overline{\mathbb{Z}}_{p}} (L(N_{ip})) \right\}$$

where  $\pi_p = \{N_{ip}\}$  is a proper system ,  $\{Z_p, \overline{Z}_p\}$  a partition of K. With the restriction to partitions which satisfy

$$(4.3) Z_p \subset \bigcup_{\mathfrak{N}_p} K_{ip}$$

one gets

#### (4.4) Theorem:

(i) Let  $\mathfrak{N}_p = \{ N_{ip} \}$  be a proper system,  $\{ \mathbf{Z}_p, \overline{\mathbf{Z}}_p \}$  a partition of K satisfying (4.3) then

(ii) Let 
$$\{z_p^i, \overline{z_p^i}\}$$
 be a partition of K , i=1,2 ,and  $z_p^1 > z_p^2$  then 
$$s(\pi_p, z_p^1, L_p) \geq s(\pi_p, z_p^2, L_p)$$

From (4.4) one can see that the best to do with respect to  $\mathfrak{N}_p$  is to choose the partition  $\{z_p^*,\overline{z_p^*}\}$  with  $z_p^*=\bigcup_{\mathfrak{N}_p}\kappa_{ip}$ , for which  $s(\mathfrak{N}_p,z_p^*,L_p)$ , additionally, has the simplified form

(4.5) 
$$S(\eta_p, Z_p^*, L_p) = E_{Z_p^*} \max_{B(\eta_p)} \{S_i + L(N_{ip})\}$$

The next question is, wether there is a possibility to improve the estimator by changing from one subproject-graph system to another. One can show

#### (4.6) Theorem:

If  $\mathcal{H}_p' = \{N_{jp}'\}$  with  $Z_p'' = \mathcal{H}_p'$   $K_{jp}', \mathcal{H}_p'' = \{N_{qp}''\}$  with  $Z_p''' = \mathcal{H}_p''$  are two proper systems with  $\mathcal{H}_p' \in \mathcal{H}_p''$  then

$$s\,\mathfrak{M}_{p}^{\prime\prime},z_{p}^{*\prime\prime},L_{p})\,\geq\,s\,(\mathfrak{N}_{p}^{\prime},z_{p}^{*\prime\prime},L_{p})$$

if  $E(\widetilde{N}_{qj}) \geq S_j - S_q$  for all  $q \in B(\mathfrak{N}_p'')$ ,  $j \in B(\mathfrak{N}_p') \cap N_{qp}''$  (where  $\widetilde{N}_{qj}$  denotes the maximal subproject-graph with respect  $q, j \in P$ )

For a proof of (4.4), (4.6) see [11].

From this representation one can get estimators known from the literature as special cases.

For instance, choosing  $S_j^1$  as PERT-estimators for the  $\widetilde{N}_{oj}$ -project passage times ,  $j \in \{0,1,\ldots,p-1\}$  ,  $\Re 1_p^1 = K(p)$  (see(2.2)) from which one gets  $B(\Re p^1) = B(p)$  ,  $Z_p^1 = \emptyset$  , gives

$$s_p^1 = S(K(p), \emptyset, L_p) = \max_{B(p)} \left\{ s_i^1 + E(X_{\phi^{-1}(i,p)}) \right\}$$
, the PERT-estimation.

Taking  $\mathfrak{N}_p^2 = \mathfrak{N}_p^1$ , but  $z_p^2 = K(p)$  and  $s_j^2 \ge s_j^1$  (where  $s_j^2$  satisfy (4.1), see [10] ) gives

$$s_{p}^{2} = s(K(p),K(p),L_{p}) = E_{K(p)} \max_{B(p)} \{ s_{i}^{2} + x_{\phi^{-1}(i,p)} \}$$

and from (4.4) it follows  $S_p^2 \ge S_p^1$ .

Choosing  $\Pi_p^3 = \{ w_{ip} \mid w_{ip} \text{ is route from i to p ,i,p} \in P \}$  in a proper way (satisfying (3.2),(3.3)) and  $S_j^3 \geqslant S_j^2$  for  $j \in B(\Pi_j^3)$  (where  $S_j^3$  satisfy (4.1),see [22]) and  $Z_p^3 = \bigcup_{n=1}^3 K(W_{ip})$  gives

$$s_{\rm p}^{3} \, = \, s\,(\{w_{\rm ip}\}, z_{\rm p}^{3}, L_{\rm p}) \, = \, E_{\rm Z_{\rm p}^{3}} \, \max_{B\,(\{w_{\rm ip}\})} \!\! \left\{ s_{\rm i}^{3} \, + \, L\,(w_{\rm ip}) \right\} \, \geqslant \, s_{\rm p}^{2}$$

because  $\mathfrak{N}_p^3$  ,  $\mathfrak{N}_p^2$  fulfill (3.5) and (4.6). At last, choosing  $\mathfrak{N}_p^4 = \left\{\widetilde{\mathtt{N}}_{op}\right\}$  ,  $z_p^4 = \widetilde{\mathtt{K}}_{op}$  , gives the special

$$S_p^4 = S(\{\widetilde{N}_{op}\}, \widetilde{K}_{op}, L_p) = E(L_p)$$

Thus, in situations in which it is difficult to determine the distribution of  $L_{m}$ , but information about the distribution of passage times with respect to subproject-graphs is available the representation given here allows the construction of estimators which acording to the amount of work one is willing to do will substantially improve all optimistic estimation possibilities known from the literature. The independence condition for the arc passage times can be weakened ( see chapter 2.) to the assumption of independence between  $X_{K_i}$ , i = 1, ..., S, for appropriate partitions  $\{K_1,\ldots,K_g\}$  of K as can be seen from the theorems. For applications additional independence between  $X_{N_{\mbox{\scriptsize ip}}}$  ,  $N_{\mbox{\scriptsize ip}} \in \mathfrak{N}_{\mbox{\scriptsize p}}$  is useful and realistic because of (3.2).

#### References

[1] Bruns,M./R.Kaerkes:Zur Einbettung strikter Halbordnungen in Netzrelationen. ORV 7(1969) 71-82. [2] Burt, J.M./D.P.Gavert/M.Perlas: Simple stochastic network: Some problems and procedures.NRLQ 4(1970)439-459. [3] Charnes, A./W.W.Cooper/G.L.Thompson: Critical path analysis via chance constrained and stochastic programming. J.ORSA 12 (1964) 460-470. [4] Clingen, C.T.: A modification of Fulkerson's PERT algorithm. J.ORSA 12(1964)629-632. [5] Elmaghraby, S.E.: On the expected duration of PERT-type networks. Manag.Sc. 13(1967)299-306. [6] Evans, J.R.: Maximum flows in probabilistic graphs-The discrete case. Networks 6(1976)161-183. [7] Fix, W. / K. Neumann / U. Steinhardt: Introduction to time and cost scheduling of decision activity networks. Discussion paper Nr.75, Inst. Wirtschaftstheorie u. Operations Research, Universität Karlsruhe (1977). [8] Frank, H.: Optimum locations on graphs with correlated normal demands J.ORSA 15(167)552-557. [9] Frank, H.: Shortest paths in probabilistic graphs. JORSA 17 (1969) 583-599. [10] Fulkerson, D.R.: Expected critical path lengths in PERT networks J.ORSA 10(1962)808-817. [11] Gaul, W.: Eine Klasse von Schätzern bei stochastischen Projektplanungsmodellen.working paper (1977). [12] Golenko,D.I.:Statistische Methoden der Netzplantechnik. Teubner Verlag Stuttgart (1972). [13] Kaerkes, R: Netzplantheorie. Proc.OR 3(1973)107-121. [14] Kaerkes, R.: Ergebnisse der Theorie der Netzpläne mit stochastisch verteilten Vorgangsdauern.Proc.OR 5(1975) 201-210. [15] Leondes, C.T./R.K.Nandi:Capacity expansion in convex cost networks with uncertain demand J.ORSA 23(1975) 1172-1178. [16] Lindsey II, J.H.: An estimate of expected critical-path

length in PERT networks.J.ORSA 20(1972) 800-812.

[17]	Lohse,D.: Eine spezielle Vorgangsdauerverteilung zur
	Netzplananalyse durch Simulation.ZOR 20(1976)115-124.
[18]	Malcolm, D.G./J.H.Roseboom/C.E.Clark/W.Fazer: Application
	of a technique for research and development program
	evaluation J.ORSA 7(1959)646-669.
[19]	Martin, J.J.: Distribution of the time through a directed
	acyclic network J.ORSA 13(1965)46-66.
[20]	Möhring, R.: Graphentheoretische Repräsentation endlicher
	Ordnungen.Proc.OR 3(1973)143-152.
[21]	Neumann, K.: Operations Research Verfahren III. Carl Hanser
	Verlag München (1975).
[22]	Robillard, P./M. Trahan: Expected completion time in PERT-
	networks.J.ORSA 24(1976)177-182.
[23]	van Slyke, R.M: Monte Carlo methods and the PERT problem.
	J.ORSA 11(1963)839-860.
[24]	van Slyke, R.M: Stochastic aspects of networks. Networks
	5(1975)97-100.
[25]	Wollmer, R.D.: Stochastic sensitivity analysis of maximum
	flows and shortest route networks. Manag.Sc. 14(1968)
	551-564.

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