Supply Chain Network Design of a Sustainable Blood Banking System

Anna Nagurney

John F. Smith Memorial Professor and

Amir H. Masoumi

Doctoral Candidate

Department of Finance and Operations Management Isenberg School of Management University of Massachusetts Amherst, Massachusetts 01003

23rd Annual POMS Conference Chicago, Illinois - April 20-23, 2012

Acknowledgments

This research was supported by the John F. Smith Memorial Fund at the University of Massachusetts Amherst. This support is gratefully acknowledged.

The authors acknowledge *Mr. Len Walker*, the Director of Business Development for the American Red Cross Blood Services in the greater Boston area, and *Dr. Jorge Rios*, the Medical Director for the American Red Cross Northeast Division Blood Services, for sharing valuable information on the subject, and enlightening thoughts on the model.

Outline

- Background and Motivation
- Some of the Relevant Literature
- The Sustainable Blood Banking System Supply Chain Network Design Model
- The Algorithm and Explicit Formulae
- Numerical Examples
- Summary and Conclusions







This talk is based on the paper:

Nagurney, A., and Masoumi, A.H. (2012), Supply Chain Network Design of a Sustainable Blood Banking System. *Sustainable Supply Chains: Models, Methods and Public Policy Implications*, Boone, T., Jayaraman, V., and Ganeshan, R., Editors, International Series in Operations Research & Management Science 174, pp49-70, Springer, London, England.

where additional background as well as references can be found.



Blood service operations are a key component of the healthcare system all over the world.



Blood service operations are a key component of the healthcare system all over the world.

A blood donation occurs when a person voluntarily has blood drawn and used for transfusions.



Blood service operations are a key component of the healthcare system all over the world.

A blood donation occurs when a person voluntarily has blood drawn and used for transfusions.

An event where donors come to give blood is called a blood drive. This can occur at a blood bank but they are often set up at a location in the community such as a shopping center, workplace, school, or house of worship.

Background and Motivation: Types of Donation

- Allogeneic (Homologous): a donor gives blood for storage at a blood bank for transfusion to an unknown recipient.
- Directed: a person, often a family member, donates blood for transfusion to a specific individual.
- Autologous: a person has blood stored that will be transfused back to the donor at a later date, usually after surgery.

Background and Motivation: Types of Donation

- Allogeneic (Homologous): a donor gives blood for storage at a blood bank for transfusion to an unknown recipient.
- Directed: a person, often a family member, donates blood for transfusion to a specific individual.
- Autologous: a person has blood stored that will be transfused back to the donor at a later date, usually after surgery.

In the developed world, most blood donors are unpaid volunteers who give blood for an established community supply. In poorer countries, donors usually give blood when family or friends need a transfusion.

Background and Motivation: Screening

Potential donors are evaluated for anything that might make their blood unsafe to use. The screening includes testing for diseases that can be transmitted by a blood transfusion, including HIV and viral hepatitis.

Background and Motivation: Screening

Potential donors are evaluated for anything that might make their blood unsafe to use. The screening includes testing for diseases that can be transmitted by a blood transfusion, including HIV and viral hepatitis.

The donor must answer questions about medical history and take a short physical examination to make sure the donation is not hazardous to his or her health.

Background and Motivation: Screening

Potential donors are evaluated for anything that might make their blood unsafe to use. The screening includes testing for diseases that can be transmitted by a blood transfusion, including HIV and viral hepatitis.

The donor must answer questions about medical history and take a short physical examination to make sure the donation is not hazardous to his or her health.

If a potential donor does not meet these criteria, they are deferred. This term is used because many donors who are ineligible may be allowed to donate later.

Whole Blood Donation:

The amount of blood drawn is typically

Whole Blood Donation:

The amount of blood drawn is typically 450-500 milliliters.

Whole Blood Donation:

The amount of blood drawn is typically 450-500 milliliters.

The blood is usually stored in a flexible plastic bag that also contains certain chemicals. This combination keeps the blood from clotting and preserves it during storage.

Whole Blood Donation:

The amount of blood drawn is typically 450-500 milliliters.

The blood is usually stored in a flexible plastic bag that also contains certain chemicals. This combination keeps the blood from clotting and preserves it during storage.

The US does not have a centralized blood donation service. The American Red Cross collects a little less than half of the blood used, the other half is collected by independent agencies, most of which are members of America's Blood Centers. The US military collects blood from service members for its own use, but also draws blood from the civilian supply.

Over 39,000 donations are needed everyday in the United States, alone, and the blood supply is frequently reported to be just 2 days away from running out (American Red Cross).

Over 39,000 donations are needed everyday in the United States, alone, and the blood supply is frequently reported to be just 2 days away from running out (American Red Cross).

Of 1,700 hospitals participating in a survey in 2007, a total of 492 reported cancellations of elective surgeries on one or more days due to blood shortages.

Over 39,000 donations are needed everyday in the United States, alone, and the blood supply is frequently reported to be just 2 days away from running out (American Red Cross).

Of 1,700 hospitals participating in a survey in 2007, a total of 492 reported cancellations of elective surgeries on one or more days due to blood shortages.

Hospitals with as many days of surgical delays as 50 or even 120 have been observed (Whitaker et al. (2007)).

Background Literature Model Examples Summary

Background and Motivation

The hospital cost of a unit of red blood cells in the US had a 6.4% increase from 2005 to 2007.

In the US, this criticality has become more of an issue in the Northeastern and Southwestern states since this cost is meaningfully higher compared to that of the Southeastern and Central states.



Life Cycles of Blood Products

The collected blood is usually stored as separate components.

- Platelets: the longest shelf life is 7 days.
- Red Blood Cells: a shelf life of 35-42 days at refrigerated temperatures
- Plasma: can be stored frozen for up to one year.

Life Cycles of Blood Products

The collected blood is usually stored as separate components.

- Platelets: the longest shelf life is 7 days.
- Red Blood Cells: a shelf life of 35-42 days at refrigerated temperatures
- Plasma: can be stored frozen for up to one year.

It is difficult to have a stockpile of blood to prepare for a disaster.

After the 9/11 terrorist attacks, it became clear that collecting during a disaster was impractical and that efforts should be focused on maintaining an adequate supply at all times.

In 2006, the national estimate for the number of units of blood components outdated by blood centers and hospitals was 1,276,000 out of 15,688,000 units

Hospitals were responsible for approximately 90% of the outdates, where this volume of medical waste imposes discarding costs to the already financially-stressed hospitals (The New York Times (2010)).



While many hospitals have their waste burned to avoid polluting the soil through landfills, the incinerators themselves are one of the nations leading sources of toxic pollutants such as dioxins and mercury (Giusti (2009)).

Healthcare facilities in the United States are second only to the food industry in producing waste, generating more than 6,600 tons per day, and more than 4 billion pounds annually (Fox News (2011)).





Some of the Relevant Literature

- Nahmias, S. (1982) Perishable inventory theory: A review.
 Operations Research 30(4), 680-708.
- Prastacos, G. P. (1984) Blood inventory management: An overview of theory and practice. Management Science 30 (7), 777-800.
- Nagurney, A., Aronson, J. (1989) A general dynamic spatial price network equilibrium model with gains and losses. Networks 19(7), 751-769.
- Pierskalla, W. P. (2004) Supply chain management of blood banks. In: Brandeau, M. L., Sanfort, F., Pierskalla, W. P., Editors, Operations Research and Health Care: A Handbook of Methods and Applications. Kluwer Academic Publishers, Boston, Massachusetts, 103-145.

- Sahin, G., Sural, H., Meral, S. (2007) Locational analysis for regionalization of Turkish Red Crescent blood services. Computers and Operations Research 34, 692-704.
- Nagurney, A., Liu, Z., Woolley, T. (2007) Sustainable supply chain and transportation networks. *International Journal of* Sustainable Transportation 1, 29-51.
- Haijema, R. (2008) Solving Large Structured Markov Decision Problems for Perishable - Inventory Management and Traffic Control, PhD thesis. Tinbergen Institute, The Netherlands.
- Mustafee, N., Katsaliaki, K., Brailsford, S.C. (2009)
 Facilitating the analysis of a UK national blood service supply chain using distributed simulation. Simulation 85(2), 113-128.

- Cetin, E., Sarul, L.S. (2009) A blood bank location model: A
 multiobjective approach. European Journal of Pure and
 Applied Mathematics 2(1), 112-124.
- Nagurney, A., Nagurney, L.S. (2010) Sustainable supply chain network design: A multicriteria perspective. *International Journal of Sustainable Engineering* 3, 189-197.
- Nagurney, A. (2010) Optimal supply chain network design and redesign at minimal total cost and with demand satisfaction. International Journal of Production Economics 128, 200-208.
- Karaesmen, I.Z., Scheller-Wolf, A., Deniz B. (2011) Managing perishable and aging inventories: Review and future research directions. In: Kempf, K. G., Kskinocak, P., Uzsoy, P., Editors, *Planning Production and Inventories in the Extended Enterprise*. Springer, Berlin, Germany, 393-436.

- Nagurney, A., Yu, M., Qiang, Q. (2011) Supply chain network design for critical needs with outsourcing. *Papers in Regional* Science 90(1), 123-143.
- Nahmias, S. (2011) Perishable Inventory Systems. Springer, New York.
- Nagurney, A., Masoumi, A.H., Yu, M. (2012) Supply chain network operations management of a blood banking system with cost and risk minimization. Computational Management Science, in press.

Background Literature Model Examples Summary

Sustainable Blood Banking System Supply Chain Network Design Model

We developed a generalized network model for design/redesign of the complex supply chain of human blood, which is a life-saving, perishable product.

More specifically, we developed a multicriteria system-optimization framework for a regionalized blood supply chain network.

Background Literature Model Examples Summary

Sustainable Blood Banking System Supply Chain Network Design Model

We assume a network topology where the top level (origin) corresponds to the organization; i.e., the regional division management of the American Red Cross. The bottom level (destination) nodes correspond to the demand sites - typically the hospitals and the other surgical medical centers.

The paths joining the origin node to the destination nodes represent sequences of supply chain network activities that ensure that the blood is collected, tested, processed, and, ultimately, delivered to the demand sites.

Components of a Regionalized Blood Banking System

ARC Regional Division Management (Top Tier),

Blood Collection

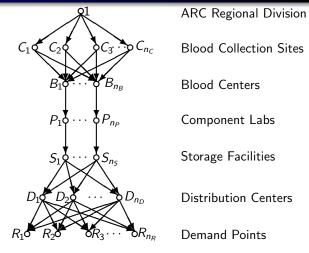
- Blood Collection Sites (Tier 2), denoted by: $C_1, C_2, ..., C_{n_C}$,

 Shipment of Collected Blood
- Blood Centers (Tier 3), denoted by: $B_1, B_2, ..., B_{n_B}$,

 Testing and Processing
- Component Labs (Tier 4), denoted by: P_1, P_2, \dots, P_{n_P} , Storage
- Storage Facilities (Tier 5), denoted by: S_1, S_2, \dots, S_{n_S} , *Shipment*
- Distribution Centers (Tier 6), denoted by: D_1, D_2, \dots, D_{n_D} ,

 Distribution
- Demand Points (Tier 7), denoted by: $R_1, R_2, \ldots, R_{n_R}$

Supply Chain Network Topology for a Regionalized Blood Bank



Graph G = [N, L], where N denotes the set of nodes and L the set of links.

Background Literature Model Examples Summary

Sustainable Blood Banking System Supply Chain Network Design Model

Our formalism is that of multicriteria optimization, where the organization seeks to determine the optimal levels of blood processed on each supply chain network link coupled with the optimal levels of capacity escalation/reduction in its blood banking supply chain network activities,

subject to:

the minimization of the total cost associated with its various activities of blood collection, shipment, processing and testing, storage, and distribution, in addition to the total discarding cost as well as the minimization of the total supply risk, subject to the uncertain demand being satisfied as closely as possible at the demand sites.

Notation

- c_a : the unit operational cost on link a.
- \hat{c}_a : the total operational cost on link a.
- f_a : the flow of whole blood/red blood cell on link a.
- p: a path in the network joining the origin node to a destination node representing the activities and their sequence.
- w_k : the pair of origin/destination (O/D) nodes $(1, R_k)$.
- \mathcal{P}_{w_k} : the set of paths, which represent alternative associated possible supply chain network processes, joining $(1, R_k)$.
- \bullet \mathcal{P} : the set of all paths joining node 1 to the demand nodes.
- n_p: the number of paths from the origin to the demand markets.
- x_p : the nonnegative flow of the blood on path p.
- d_k : the uncertain demand for blood at demand location k.
- v_k : the projected demand for blood at demand location k.

Formulation

Total Operational Cost on Link a

$$\hat{c}_a(f_a) = f_a \times c_a(f_a), \qquad \forall a \in L, \tag{1}$$

assumed to be convex and continuously differentiable.

Let P_k be the probability distribution function of d_k , that is, $P_k(D_k) = P_k(d_k \leq D_k) = \int_0^{D_k} \mathcal{F}_k(t) d(t)$. Therefore,

Shortage and Surplus of Blood at Demand Point R_k

$$\Delta_k^- \equiv \max\{0, d_k - v_k\}, \qquad k = 1, \dots, n_R, \tag{2}$$

$$\Delta_k^+ \equiv \max\{0, v_k - d_k\}, \qquad k = 1, \dots, n_R, \tag{3}$$

Expected Values of Shortage and Surplus

$$E(\Delta_k^-) = \int_{\nu_k}^{\infty} (t - \nu_k) \mathcal{F}_k(t) d(t), \qquad k = 1, \dots, n_R, \quad (4)$$

$$E(\Delta_k^+) = \int_0^{\nu_k} (\nu_k - t) \mathcal{F}_k(t) d(t), \qquad k = 1, \dots, n_R.$$
 (5)

Expected Total Penalty at Demand Point k

$$E(\lambda_k^- \Delta_k^- + \lambda_k^+ \Delta_k^+) = \lambda_k^- E(\Delta_k^-) + \lambda_k^+ E(\Delta_k^+), \tag{6}$$

where λ_k^- is a large penalty associated with the shortage of a unit of blood, and λ_k^+ is the incurred cost of a unit of surplus blood.

Formulation

Arc Multiplier, and Waste/Loss on a link

Let α_a correspond to the percentage of loss over link a, and f_a' denote the final flow on that link. Thus,

$$f_a' = \alpha_a f_a, \quad \forall a \in L.$$
 (7)

Therefore, the waste/loss on link a, denoted by w_a , is equal to:

$$w_a = f_a - f_a' = (1 - \alpha_a)f_a, \qquad \forall a \in L. \tag{8}$$

Total Discarding Cost function

$$\hat{z}_a = \hat{z}_a(f_a), \qquad \forall a \in L.$$
 (9)

Non-negativity of Flows

$$x_p \ge 0, \qquad \forall p \in \mathcal{P},$$
 (10)

Path Multiplier, and Projected Demand

$$\mu_{p} \equiv \prod_{\mathbf{a} \in p} \alpha_{\mathbf{a}}, \qquad \forall p \in \mathcal{P},$$
(11)

where μ_p is the throughput factor on path p. Thus, the projected demand at R_k is equal to:

$$v_k \equiv \sum_{p \in \mathcal{P}_{w_k}} x_p \mu_p, \qquad k = 1, \dots, n_R.$$
 (12)

Relation between Link and Path Flows

$$\alpha_{ap} \equiv \begin{cases} \delta_{ap} \prod_{a' < a} \alpha_{a'}, & \text{if } \{a' < a\} \neq \emptyset, \\ \delta_{ap}, & \text{if } \{a' < a\} = \emptyset, \end{cases}$$

$$(13)$$

where $\{a' < a\}$ denotes the set of the links preceding link a in path p. Also, δ_{ap} is defined as equal to 1 if link a is contained in path p; otherwise, it is equal to zero. Therefore,

$$f_a = \sum_{p \in \mathcal{P}} x_p \ \alpha_{ap}, \qquad \forall a \in L.$$
 (14)

Total Investment Cost of Capacity Enhancement/Reduction on Links

$$\hat{\pi}_{\mathsf{a}} = \hat{\pi}_{\mathsf{a}}(u_{\mathsf{a}}),\tag{15}$$

where u_a denotes the change in capacity on link a, and $\hat{\pi}_a$ is the total investment cost of such change.

Capacity Adjustments Constraints

$$f_a \le \bar{u}_a + u_a, \qquad \forall a \in L,$$
 (16)

and

$$-\bar{u}_a \le u_a, \qquad \forall a \in L,$$
 (17)

where \bar{u}_a denotes the nonnegative existing capacity on link a.

Cost Objective Function

Minimization of Total Costs

Minimize
$$\sum_{a \in L} \hat{c}_a(f_a) + \sum_{a \in L} \hat{z}_a(f_a) + \sum_{a \in L} \hat{\pi}_a(u_a)$$
$$+ \sum_{k=1}^{n_R} (\lambda_k^- E(\Delta_k^-) + \lambda_k^+ E(\Delta_k^+)), \tag{18}$$

subject to: constraints (10), (12), (14), (16), and (17).

Supply Side Risk

One of the most significant challenges for the ARC is to capture the risk associated with different activities in the blood supply chain network. Unlike the demand which can be projected, albeit with some uncertainty involved, the amount of donated blood at the collection sites has been observed to be highly stochastic.

Risk Objective Function

Minimize
$$\sum_{a \in L_1} \hat{r}_a(f_a)$$
, (19)

where $\hat{r}_a = \hat{r}_a(f_a)$ is the total risk function on link a, and L_1 is the set of blood collection links.

The Multicriteria Optimization Formulation

 θ : the weight associated with the risk objective function, assigned by the decision maker.

Multicriteria Optimization Formulation in Terms of Link Flows

Minimize
$$\sum_{a \in L} \hat{c}_a(f_a) + \sum_{a \in L} \hat{z}_a(f_a) + \sum_{a \in L} \hat{\pi}_a(u_a)$$

$$+\sum_{k=1}^{n_R} \left(\lambda_k^- E(\Delta_k^-) + \lambda_k^+ E(\Delta_k^+)\right) + \theta \sum_{a \in L_1} \hat{r}_a(f_a), \qquad (20)$$

subject to: constraints (10), (12), (14), (16), and (17).

The Multicriteria Optimization Formulation

Multicriteria Optimization Formulation in Terms of Path Flows

Minimize
$$\sum_{p \in \mathcal{P}} (\hat{C}_p(x) + \hat{Z}_p(x)) + \sum_{a \in L} \hat{\pi}_a(u_a)$$

$$+ \sum_{k=1}^{n_R} \left(\lambda_k^- E(\Delta_k^-) + \lambda_k^+ E(\Delta_k^+) \right) + \theta \sum_{\rho \in \mathcal{P}} \hat{R}_{\rho}(x), \qquad (21)$$

subject to: constraints (10), (12), (16), and (17).

The total costs on path p are expressed as:

$$\hat{C}_p(x) = x_p \times C_p(x), \qquad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R,$$
 (22a)

$$\hat{Z}_p(x) = x_p \times Z_p(x), \qquad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R,$$
 (22b)

$$\hat{R}_p(x) = x_p \times R_p(x), \quad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R,$$
 (22c)

The unit cost functions on path p are, in turn, defined as below:

$$C_p(x) \equiv \sum_{a \in I} c_a(f_a) \alpha_{ap}, \quad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R,$$
 (23a)

$$Z_p(x) \equiv \sum_{a \in L} z_a(f_a) \alpha_{ap}, \qquad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R,$$
 (23b)

$$R_p(x) \equiv \sum_{a \in L_1} r_a(f_a) \alpha_{ap}, \qquad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R.$$
 (23c)

Formulation: Preliminaries

It is proved that the partial derivatives of expected shortage at the demand locations with respect to the path flows are derived from:

$$\frac{\partial E(\Delta_{k}^{-})}{\partial x_{p}} = \mu_{p} \left[P_{k} \left(\sum_{p \in \mathcal{P}_{w_{k}}} x_{p} \mu_{p} \right) - 1 \right], \ \forall p \in \mathcal{P}_{w_{k}}; k = 1, \dots, n_{R}.$$
(24a)

Similarly, for surplus we have:

$$\frac{\partial E(\Delta_{k}^{+})}{\partial x_{p}} = \mu_{p} P_{k} \left(\sum_{p \in \mathcal{P}_{w_{k}}} x_{p} \mu_{p} \right), \ \forall p \in \mathcal{P}_{w_{k}}; k = 1, \dots, n_{R}.$$
(24b)

Formulation: Lemma

The following lemma was developed to help us calculate the partial derivatives of the cost functions:

Lemma 1

$$\frac{\partial (\sum_{q \in \mathcal{P}} \hat{C}_q(x))}{\partial x_p} \equiv \sum_{a \in L} \frac{\partial \hat{c}_a(f_a)}{\partial f_a} \alpha_{ap}, \quad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R, \quad (25a)$$

$$\frac{\partial (\sum_{q \in \mathcal{P}} \hat{Z}_q(x))}{\partial x_p} \equiv \sum_{a \in L} \frac{\partial \hat{z}_a(f_a)}{\partial f_a} \alpha_{ap}, \quad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R, \quad (25b)$$

$$\frac{\partial (\sum_{q \in \mathcal{P}} \hat{R}_q(x))}{\partial x_p} \equiv \sum_{a \in L_1} \frac{\partial \hat{r}_a(f_a)}{\partial f_a} \alpha_{ap}, \quad \forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R. \quad (25c)$$

Variational Inequality Formulation: Feasible Set and Decision Variables

Let K denote the feasible set such that:

$$K \equiv \{(x, u, \gamma) | x \in R_+^{n_p}, (17) \text{ holds, and } \gamma \in R_+^{n_L} \}.$$
 (26)

Our multicriteria optimization problem is characterized, under our assumptions, by a convex objective function and a convex feasible set.

We group the path flows, the link flows, and the projected demands into the respective vectors x, f, and v. Also, the link capacity changes are grouped into the vector u. Lastly, the Lagrange multipliers corresponding to the links capacity adjustment constraints are grouped into the vector γ .

Variational Inequality Formulation (in Term of Path Flows)

Theorem 1

Our multicriteria optimization problem, subject to its constraints, is equivalent to the variational inequality problem: determine the vector of optimal path flows, the vector of optimal capacity adjustments, and the vector of optimal Lagrange multipliers $(x^*, u^*, \gamma^*) \in K$, such that:

$$\sum_{k=1}^{n_R} \sum_{p \in \mathcal{P}_{w_k}} \left[\frac{\partial (\sum_{q \in \mathcal{P}} \hat{C}_q(x^*))}{\partial x_p} + \frac{\partial (\sum_{q \in \mathcal{P}} \hat{Z}_q(x^*))}{\partial x_p} + \lambda_k^+ \mu_p P_k \left(\sum_{p \in \mathcal{P}_{w_k}} x_p^* \mu_p \right) \right] \\
- \lambda_k^- \mu_p \left(1 - P_k \left(\sum_{p \in \mathcal{P}_{w_k}} x_p^* \mu_p \right) \right) + \sum_{a \in L} \gamma_a^* \delta_{ap} + \theta \left[\frac{\partial (\sum_{q \in \mathcal{P}} \hat{R}_q(x^*))}{\partial x_p} \right] \times [x_p - x_p^*] \\
+ \sum_{a \in L} \left[\frac{\partial \hat{\pi}_a(u_a^*)}{\partial u_a} - \gamma_a^* \right] \times [u_a - u_a^*] + \sum_{a \in L} \left[\bar{u}_a + u_a^* - \sum_{p \in \mathcal{P}} x_p^* \alpha_{ap} \right] \times [\gamma_a - \gamma_a^*] \ge 0, \\
\forall (x, u, \gamma) \in \mathcal{K}. \tag{27}$$

Variational Inequality Formulation (in Terms of Link Flows)

Theorem 1 (cont'd)

The variational inequality mentioned, in turn, can be rewritten in terms of link flows as: determine the vector of optimal link flows, the vectors of optimal projected demands and the link capacity adjustments, and the vector of optimal Lagrange multipliers $(f^*, v^*, u^*, \gamma^*) \in K^1$, such that:

$$\sum_{a \in I} \left[\frac{\partial \hat{c}_{a}(f_{a}^{*})}{\partial f_{a}} + \frac{\partial \hat{z}_{a}(f_{a}^{*})}{\partial f_{a}} + \gamma_{a}^{*} + \theta \left[\frac{\partial \hat{r}_{a}(f_{a}^{*})}{\partial f_{a}} \right] \times [f_{a} - f_{a}^{*}] \right]$$

$$+ \sum_{a \in L} \left[\frac{\partial \hat{\pi}_{a}(u_{a}^{*})}{\partial u_{a}} - \gamma_{a}^{*} \right] \times \left[u_{a} - u_{a}^{*} \right] + \sum_{k=1}^{n_{R}} \left[\lambda_{k}^{+} P_{k}(v_{k}^{*}) - \lambda_{k}^{-} (1 - P_{k}(v_{k}^{*})) \right] \times \left[v_{k} - v_{k}^{*} \right]$$

$$+\sum_{a\in I}\left[\bar{u}_a+u_a^*-f_a^*\right]\times\left[\gamma_a-\gamma_a^*\right]\geq 0, \qquad \forall (f,v,u,\gamma)\in K^1, \tag{28}$$

where K^1 denotes the feasible set as defined below:

$$K^1 \equiv \{(f, v, u, \gamma) | \exists x \ge 0, (12), (14), \text{ and } (17) \text{ hold, and } \gamma \ge 0\}.$$
 (29)

The Solution Algorithm

The realization of Euler Method for the solution of the sustainable blood bank supply chain network design problem governed by the developed variational inequalities induces subproblems that can be solved explicitly and in closed form.

At iteration τ of the Euler method one computes:

$$X^{\tau+1} = P_{\mathcal{K}}(X^{\tau} - a_{\tau}F(X^{\tau})), \tag{30}$$

where P_K is the projection on the feasible set K, and F is the function that enters the standard form variational inequality problem.

Explicit Formulae for the Euler Method Applied to Our Variational Inequality Formulation

$$\begin{split} & x_{\rho}^{\tau+1} = \max\{0, x_{\rho}^{\tau} + a_{\tau}(\lambda_{k}^{-}\mu_{\rho}(1 - P_{k}(\sum_{\rho \in \mathcal{P}_{w_{k}}} x_{\rho}^{\tau}\mu_{\rho})) - \lambda_{k}^{+}\mu_{\rho}P_{k}(\sum_{\rho \in \mathcal{P}_{w_{k}}} x_{\rho}^{\tau}\mu_{\rho}) \\ & - \frac{\partial(\sum_{q \in \mathcal{P}} \hat{C}_{q}(x^{\tau}))}{\partial x_{\rho}} - \frac{\partial(\sum_{q \in \mathcal{P}} \hat{Z}_{q}(x^{\tau}))}{\partial x_{\rho}} - \sum_{r \in I} \gamma_{a}^{\tau} \delta_{a\rho} - \theta \frac{\partial(\sum_{q \in \mathcal{P}} \hat{R}_{q}(x^{\tau}))}{\partial x_{\rho}})\}, \end{split}$$

$$\forall p \in \mathcal{P}_{w_k}; k = 1, \dots, n_R; \tag{31a}$$

$$u_{a}^{\tau+1} = \max\{-\bar{u}_{a}, u_{a}^{\tau} + a_{\tau}(\gamma_{a}^{\tau} - \frac{\partial \hat{\pi}_{a}(u_{a}^{\tau})}{\partial u_{a}})\}, \quad \forall a \in L;$$
 (31b)

$$\gamma_{a}^{\tau+1} = \max\{0, \gamma_{a}^{\tau} + a_{\tau}(\sum_{p \in \mathcal{P}} x_{p}^{\tau} \alpha_{ap} - \bar{u}_{a} - u_{a}^{\tau})\}, \quad \forall a \in L.$$
 (31c)

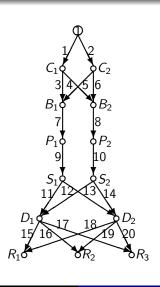
We applied the above to calculate updated product flows, capacity changes, and Lagrange multipliers during the steps of the Euler Method.

Background Literature Model Examples Summary

Illustrative Numerical Examples



The Supply Chain Network Topology for the Numerical Examples



ARC Regional Division

Blood Collection Sites

Blood Centers

Component Labs

Storage Facilities

Distribution Centers

Demand Points

Example 1: Design from Scratch

The demands at these demand points followed uniform probability distribution on the intervals [5,10], [40,50], and [25,40], respectively:

$$P_{1}\left(\sum_{p\in\mathcal{P}_{w_{1}}}\mu_{p}x_{p}\right) = \frac{\sum_{p\in\mathcal{P}_{w_{1}}}\mu_{p}x_{p} - 5}{5}, \ P_{2}\left(\sum_{p\in\mathcal{P}_{w_{2}}}\mu_{p}x_{p}\right) = \frac{\sum_{p\in\mathcal{P}_{w_{2}}}\mu_{p}x_{p} - 40}{10},$$

$$P_{3}\left(\sum_{p\in\mathcal{P}_{w_{3}}}\mu_{p}x_{p}\right) = \frac{\sum_{p\in\mathcal{P}_{w_{3}}}\mu_{p}x_{p} - 25}{15}.$$

$$\lambda_{1}^{-} = 2800, \quad \lambda_{1}^{+} = 50,$$

$$\lambda_{2}^{-} = 3000, \quad \lambda_{2}^{+} = 60,$$

$$\lambda_{3}^{-} = 3100, \quad \lambda_{3}^{+} = 50.$$

$$\hat{r}_{1}(f_{1}) = 2f_{1}^{2}, \ \hat{r}_{2}(f_{2}) = 1.5f_{2}^{2}, \ \text{and} \ \theta = 0.7$$

Solution Procedure

The Euler method for the solution of variational inequality (27) was implemented in Matlab. A Microsoft Windows System with a Dell PC at the University of Massachusetts Amherst was used for all the computations. We set the sequence $a_{\tau} = .1(1, \frac{1}{2}, \frac{1}{2}, \cdots)$, and the convergence tolerance was $\epsilon = 10^{-6}$.

Background Literature Model Examples Summary

Example 1 Results: Total Cost Functions and Solution

Link a	α_{a}	$\hat{c}_a(f_a)$	$\hat{z}_a(f_a)$	$\hat{\pi}_a(u_a)$	f_a^*	u _a *	γ_a^*
1	.97	$6f_1^2 + 15f_1$	$.8f_1^2$	$.8u_1^2 + u_1$	47.18	47.18	76.49
2	.99	$9f_2^2 + 11f_2$	$.7f_2^2$	$.6u_2^2 + u_2$	39.78	39.78	48.73
3	1.00	$.7f_3^2 + f_3$	$.6f_3^2$	$u_3^2 + 2u_3$	25.93	25.93	53.86
4	.99	$1.2f_4^2 + f_4$	$.8f_4^2$	$2u_4^2 + u_4$	19.38	19.38	78.51
5	1.00	$f_5^2 + 3f_5$	$.6f_5^2$	$u_5^2 + u_5$	18.25	18.25	37.50
6	1.00	$.8f_6^2 + 2f_6$	$.8f_6^2$	$1.5u_6^2 + 3u_6$	20.74	20.74	65.22
7	.92	$2.5f_7^2 + 2f_7$	$.5f_7^2$	$7u_7^2 + 12u_7$	43.92	43.92	626.73
8	.96	$3f_8^2 + 5f_8$	$.8f_8^2$	$6u_8^2 + 20u_8$	36.73	36.73	460.69
9	.98	$.8f_9^2 + 6f_9$	$.4f_9^2$	$3u_9^2 + 2u_9$	38.79	38.79	234.74
10	1.00	$.5f_{10}^2 + 3f_{10}$	$.7f_{10}^2$	$5.4u_{10}^2 + 2u_{10}$	34.56	34.56	375.18
11	1.00	$.3f_{11}^2 + f_{11}$	$.3f_{11}^2$	$u_{11}^2 + u_{11}$	25.90	25.90	52.80
12	1.00	$.5f_{12}^2 + 2f_{12}$	$.4f_{12}^2$	$1.5u_{12}^2 + u_{12}$	12.11	12.11	37.34
13	1.00	$.4f_{13}^2 + 2f_{13}$	$.3f_{13}^2$	$1.8u_{13}^2 + 1.5u_{13}$	17.62	17.62	64.92
14	1.00	$.6f_{14}^2 + f_{14}$	$.4f_{14}^2$	$u_{14}^2 + 2u_{14}$	16.94	16.94	35.88
15	1.00	$.4f_{15}^2 + f_{15}$	$.7f_{15}^2$	$.5u_{15}^2 + 1.1u_{15}$	5.06	5.06	6.16
16	1.00	$.8f_{16}^2 + 2f_{16}$	$.4f_{16}^{2}$	$.7u_{16}^2 + 3u_{16}$	24.54	24.54	37.36
17	.98	$.5f_{17}^2 + 3f_{17}$	$.5f_{17}^2$	$2u_{17}^2 + u_{17}$	13.92	13.92	56.66
18	1.00	$.7f_{18}^2 + f_{18}$	$.7f_{18}^2$	$u_{18}^2 + u_{18}$	0.00	0.00	1.00
19	1.00	$.6f_{19}^2 + 4f_{19}$	$.4f_{19}^2$	$u_{19}^2 + 2u_{19}$	15.93	15.93	33.86
20	.98	$1.1f_{20}^2 + 5f_{20}$	$.5f_{20}^{2}$	$.8u_{20}^2 + u_{20}$	12.54	12.54	21.06

Example 1 Results (cont'd)

The values of the total investment cost and the cost objective criterion were 42, 375.96 and 135, 486.43, respectively.

The computed amounts of projected demand for each of the three demand points were:

$$v_1^* = 5.06$$
, $v_2^* = 40.48$, and $v_3^* = 25.93$.

Note that the values of the projected demand were closer to the lower bounds of their uniform probability distributions due to the relatively high cost of setting up a new blood supply chain network from scratch.

Example 2: Increased Penalties

Example 2 had the exact same data as Example 1 with the exception of the penalties per unit shortage which were ten times larger.

$$\lambda_1^- = 28000, \qquad \lambda_2^- = 30000, \qquad \lambda_3^- = 31000.$$

Background Literature Model Examples Summary

Example 2 Results: Total Cost Functions and Solution

Link a	α_{a}	$\hat{c}_a(f_a)$	$\hat{z}_a(f_a)$	$\hat{\pi}_a(u_a)$	f_a^*	u _a *	γ_a^*
1	.97	$6f_1^2 + 15f_1$	$.8f_1^2$	$.8u_1^2 + u_1$	63.53	63.53	102.65
2	.99	$9f_2^2 + 11f_2$	$.7f_2^2$	$.6u_2^2 + u_2$	53.53	53.53	65.23
3	1.00	$-7f_3^2 + f_3$	$.6f_3^2$	$u_3^2 + 2u_3$	34.93	34.93	71.85
4	.99	$1.2f_4^2 + f_4$	$.8f_4^2$	$2u_4^2 + u_4$	26.08	26.08	105.34
5	1.00	$f_5^2 + 3f_5$	$.6f_5^2$	$u_5^2 + u_5$	24.50	24.50	50.00
6	1.00	$.8f_6^2 + 2f_6$	$.8f_6^2$	$1.5u_6^2 + 3u_6$	27.96	27.96	86.89
7	.92	$2.5f_7^2 + 2f_7$	$.5f_7^2$	$7u_7^2 + 12u_7$	59.08	59.08	839.28
8	.96	$3f_8^2 + 5f_8$	$.8f_8^2$	$6u_8^2 + 20u_8$	49.48	49.48	613.92
9	.98	$.8f_9^2 + 6f_9$	$.4f_9^2$	$3u_9^2 + 2u_9$	52.18	52.18	315.05
10	1.00	$.5f_{10}^2 + 3f_{10}$	$.7f_{10}^2$	$5.4u_{10}^2 + 2u_{10}$	46.55	46.55	504.85
11	1.00	$.3f_{11}^2 + f_{11}$	$.3f_{11}^2$	$u_{11}^2 + u_{11}$	35.01	35.01	71.03
12	1.00	$.5f_{12}^2 + 2f_{12}$	$.4f_{12}^2$	$1.5u_{12}^2 + u_{12}$	16.12	16.12	49.36
13	1.00	$.4f_{13}^2 + 2f_{13}$	$.3f_{13}^2$	$1.8u_{13}^2 + 1.5u_{13}$	23.93	23.93	87.64
14	1.00	$.6f_{14}^2 + f_{14}$	$.4f_{14}^2$	$u_{14}^2 + 2u_{14}$	22.63	22.63	47.25
15	1.00	$.4f_{15}^2 + f_{15}$	$.7f_{15}^2$	$.5u_{15}^2 + 1.1u_{15}$	9.33	9.33	10.43
16	1.00	$.8f_{16}^2 + 2f_{16}$	$.4f_{16}^{2}$	$.7u_{16}^2 + 3u_{16}$	29.73	29.73	44.62
17	.98	$.5f_{17}^2 + 3f_{17}$	$.5f_{17}^2$	$2u_{17}^2 + u_{17}$	19.89	19.89	80.55
18	1.00	$.7f_{18}^2 + f_{18}$	$.7f_{18}^2$	$u_{18}^2 + u_{18}$	0.00	0.00	1.00
19	1.00	$.6f_{19}^2 + 4f_{19}$	$.4f_{19}^2$	$u_{19}^2 + 2u_{19}$	18.99	18.99	39.97
20	.98	$1.1f_{20}^2 + 5f_{20}$	$.5f_{20}^{2}$	$.8u_{20}^2 + u_{20}$	18.98	18.98	31.37

Example 2 Results (cont'd)

Raising the shortage penalties increased the level of activities in almost all the network links. The new projected demand values were:

$$v_1^* = 9.33$$
, $v_2^* = 48.71$, and $v_3^* = 38.09$.

Here the projected demand values were closer to the upper bounds of their uniform probability distributions.

Thus, the values of the total investment cost and the cost objective criterion, were 75,814.03 and 177,327.31, respectively, which were significantly higher than Example 1.

Example 3: Redesign Problem

The existing capacities for links were chosen close to the optimal solution for corresponding capacities in Example 1.

All other parameters were the same as in Example 1.

Background Literature Model Examples Summary

Example 3 Results: Total Cost Functions and Solution

Link a	α_{a}	$\hat{c}_a(f_a)$	$\hat{z}_a(f_a)$	$\hat{\pi}_a(u_a)$	ūа	f_a^*	u _a *	γ_a^*
1	.97	$6f_1^2 + 15f_1$	$.8f_1^2$	$.8u_1^2 + u_1$	48	54.14	6.14	10.83
2	.99	$9f_2^2 + 11f_2$	$.7f_2^2$	$.6u_2^2 + u_2$	40	43.85	3.85	5.62
3	1.00	$.7f_3^2 + f_3$	$.6f_3^2$	$u_3^2 + 2u_3$	26	29.64	3.64	9.29
4	.99	$1.2f_4^2 + f_4$	$.8f_4^2$	$2u_4^2 + u_4$	20	22.35	2.35	10.39
5	1.00	$f_5^2 + 3f_5$	$.6f_5^2$	$u_5^2 + u_5$	19	20.10	1.10	3.20
6	1.00	$.8f_6^2 + 2f_6$	$.8f_6^2$	$1.5u_6^2 + 3u_6$	21	22.88	1.88	8.63
7	.92	$2.5f_7^2 + 2f_7$	$.5f_7^2$	$7u_7^2 + 12u_7$	44	49.45	5.45	88.41
8	.96	$3f_8^2 + 5f_8$	$.8f_8^2$ $.4f_9^2$	$6u_8^2 + 20u_8$	37	41.40	4.40	72.88
9	.98	$.8f_9^2 + 6f_9$	$.4f_9^2$	$3u_9^2 + 2u_9$	39	43.67	4.67	30.04
10	1.00	$.5f_{10}^2 + 3f_{10}$	$.7f_{10}^2$	$5.4u_{10}^2 + 2u_{10}$	35	38.95	3.95	44.70
11	1.00	$.3f_{11}^2 + f_{11}$	$.3f_{11}^2$	$u_{11}^2 + u_{11}$	26	29.23	3.23	7.45
12	1.00	$.5f_{12}^2 + 2f_{12}$	$.4f_{12}^{2}$	$1.5u_{12}^2 + u_{12}$	13	13.57	0.57	2.72
13	1.00	$.4f_{13}^2 + 2f_{13}$	$.3f_{13}^2$	$1.8u_{13}^2 + 1.5u_{13}$	18	22.05	4.05	16.07
14	1.00	$.6f_{14}^2 + f_{14}$	$.4f_{14}^2$ $.7f_{15}^2$	$u_{14}^2 + 2u_{14}$	17	16.90	-0.10	1.81
15	1.00	$.4f_{15}^2 + f_{15}$	$.7f_{15}^2$	$.5u_{15}^2 + 1.1u_{15}$	6	6.62	0.62	1.72
16	1.00	$.8f_{16}^2 + 2f_{16}$	$.4f_{16}^2$	$.7u_{16}^2 + 3u_{16}$	25	25.73	0.73	4.03
17	.98	$.5f_{17}^2 + 3f_{17}$	$.5f_{17}^2$	$2u_{17}^2 + u_{17}$	14	18.92	4.92	20.69
18	1.00	$.7f_{18}^2 + f_{18}$	$.7f_{18}^2$	$u_{18}^2 + u_{18}$	0	0.00	0.00	1.00
19	1.00	$.6f_{19}^2 + 4f_{19}$	$.4f_{19}^2$	$u_{19}^2 + 2u_{19}$	16	17.77	1.77	5.53
20	.98	$1.1f_{20}^2 + 5f_{20}$	$.4f_{19}^{2}$ $.5f_{20}^{2}$	$.8u_{20}^2 + u_{20}$	13	12.10	-0.62	0.00

Example 3 Results (cont'd)

The optimal Lagrangian multipliers, γ_a^* , i.e., the shadow prices of the capacity adjustment constraints, were considerably smaller than their counterparts in Example 1.

So, the respective values of the capacity investment cost and the cost criterion were 856.36 and 85, 738.13.

The computed projected demand values:

$$v_1^* = 6.62$$
, $v_2^* = 43.50$, and $v_3^* = 30.40$.

Example 4: Increased Demands

The existing capacities, the shortage penalties, and the cost functions were the same as in Example 3.

However, the demands at the three hospitals were escalated, following uniform probability distributions on the intervals [10,17], [50,70], and [30,60], respectively.

Background Literature Model Examples Summary

Example 4 Results: Total Cost Functions and Solution

Link a	α_{a}	$\hat{c}_a(f_a)$	$\hat{z}_a(f_a)$	$\hat{\pi}_a(u_a)$	ūa	f_a^*	u _a *	γ_a^*
1	.97	$6f_1^2 + 15f_1$	$.8f_1^2$	$.8u_1^2 + u_1$	48	65.45	17.45	28.92
2	.99	$9f_2^2 + 11f_2$	$.7f_2^2$	$.6u_2^2 + u_2$	40	53.36	13.36	17.03
3	1.00	$.7f_3^2 + f_3$	$.6f_3^2$	$u_3^2 + 2u_3$	26	35.87	9.87	21.74
4	.99	$1.2f_4^2 + f_4$	$.8f_4^2$	$2u_4^2 + u_4$	20	26.98	6.98	28.91
5	1.00	$f_5^2 + 3f_5$	$.6f_5^2$	$u_5^2 + u_5$	19	24.43	5.43	11.86
6	1.00	$.8f_6^2 + 2f_6$	$.8f_6^2$	$1.5u_6^2 + 3u_6$	21	27.87	6.87	23.60
7	.92	$2.5f_7^2 + 2f_7$	$.5f_7^2$	$7u_7^2 + 12u_7$	44	59.94	15.94	234.92
8	.96	$3f_8^2 + 5f_8$	$.8f_8^2$	$6u_8^2 + 20u_8$	37	50.21	13.21	178.39
9	.98	$.8f_9^2 + 6f_9$	$.4f_9^2$	$3u_9^2 + 2u_9$	39	52.94	13.94	85.77
10	1.00	$.5f_{10}^2 + 3f_{10}$	$.7f_{10}^2$	$5.4u_{10}^2 + 2u_{10}$	35	47.24	12.24	134.64
11	1.00	$.3f_{11}^2 + f_{11}$	$.3f_{11}^{2}$	$u_{11}^2 + u_{11}$	26	35.68	9.68	20.35
12	1.00	$.5f_{12}^{2} + 2f_{12}$	$.4f_{12}^{2}$	$1.5u_{12}^2 + u_{12}$	13	16.20	3.20	10.61
13	1.00	$.4f_{13}^2 + 2f_{13}$	$.3f_{13}^{2}$	$1.8u_{13}^2 + 1.5u_{13}$	18	26.54	8.54	32.23
14	1.00	$.6f_{14}^2 + f_{14}$	$.4f_{14}^{2}$ $.7f_{15}^{2}$	$u_{14}^2 + 2u_{14}$	17	20.70	3.70	9.40
15	1.00	$.4f_{15}^2 + f_{15}$	$.7f_{15}^2$	$.5u_{15}^2 + 1.1u_{15}$	6	10.30	4.30	5.40
16	1.00	$.8f_{16}^2 + 2f_{16}$	$.4f_{16}^2$	$.7u_{16}^2 + 3u_{16}$	25	30.96	5.96	11.34
17	.98	$.5f_{17}^2 + 3f_{17}$	$.5f_{17}^2$	$2u_{17}^2 + u_{17}$	14	20.95	6.95	28.81
18	1.00	$.7f_{18}^2 + f_{18}$	$.7f_{18}^2$	$u_{18}^2 + u_{18}$	0	0.35	0.35	1.69
19	1.00	$.6f_{19}^2 + 4f_{19}$	$.4f_{19}^2$	$u_{10}^2 + 2u_{19}$	16	21.68	5.68	13.36
20	.98	$1.1\overline{f_{20}^2} + 5f_{20}$	$.4f_{19}^{2}$ $.5f_{20}^{2}$	$.8u_{20}^2 + u_{20}$	13	14.14	1.14	2.83

Example 4 Results (cont'd)

A 50% increase in demand resulted in significant positive capacity changes as well as positive flows on all 20 links in the network.

The values of the total investment function and the cost criterion were 5, 949.18 and 166, 445.73, respectively.

The projected demand values were now:

$$v_1^* = 10.65$$
, $v_2^* = 52.64$, and $v_3^* = 34.39$.

Example 5: Decreased Demands

Example 5 was similar to Example 4, but now the demand suffered a decrease from the original demand scenario.

The demand at demand points 1, 2, and 3 followed a uniform probability distribution on the intervals [4,7], [30,40], and [15,30], respectively.

Background Literature Model Examples Summary

Example 5 Results: Total Cost Functions and Solution

Link a	α_{a}	$\hat{c}_a(f_a)$	$\hat{z}_a(f_a)$	$\hat{\pi}_a(u_a)$	ū _a	f_a^*	u _a *	γ_a^*
1	.97	$6f_1^2 + 15f_1$	$.8f_1^2$	$.8u_1^2 + u_1$	48	43.02	-0.62	0.00
2	.99	$9f_2^2 + 11f_2$	$.7f_2^2$	$.6u_2^2 + u_2$	40	34.54	-0.83	0.00
3	1.00	$.7f_3^2 + f_3$	$.6f_3^2$	$u_3^2 + 2u_3$	26	23.77	-1.00	0.00
4	.99	$1.2f_4^2 + f_4$	$.8f_4^2$	$2u_4^2 + u_4$	20	17.54	-0.25	0.00
5	1.00	$f_5^2 + 3f_5$	$.8f_4^2$ $.6f_5^2$	$u_5^2 + u_5$	19	15.45	-0.50	0.00
6	1.00	$.8f_6^2 + 2f_6$	$.8f_6^2$	$1.5u_6^2 + 3u_6$	21	18.40	-1.00	0.00
7	.92	$2.5f_7^2 + 2f_7$	$.5f_7^2$	$7u_7^2 + 12u_7$	44	38.99	-0.86	0.00
8	.96	$3f_8^2 + 5f_8$	$.8f_8^2$	$6u_8^2 + 20u_8$	37	32.91	-1.67	0.00
9	.98	$.8f_9^2 + 6f_9$	$.4f_{0}^{2}$	$3u_9^2 + 2u_9$	39	34.43	-0.33	0.00
10	1.00	$.5f_{10}^2 + 3f_{10}$	$.7f_{10}^2$	$5.4u_{10}^2 + 2u_{10}$	35	30.96	-0.19	0.00
11	1.00	$.3f_{11}^2 + f_{11}$	$.3f_{11}^2$	$u_{11}^2 + u_{11}$	26	23.49	-0.50	0.00
12	1.00	$.5f_{12}^2 + 2f_{12}$	$.4f_{12}^{2}$	$1.5u_{12}^2 + u_{12}$	13	10.25	-0.33	0.00
13	1.00	$.4f_{13}^2 + 2f_{13}$	$.3f_{13}^2$	$1.8u_{13}^2 + 1.5u_{13}$	18	18.85	0.85	4.57
14	1.00	$.6f_{14}^2 + f_{14}$	$.4f_{14}^2$	$u_{14}^2 + 2u_{14}$	17	12.11	-1.00	0.00
15	1.00	$.4f_{15}^2 + f_{15}$	$.7f_{15}^2$	$.5u_{15}^2 + 1.1u_{15}$	6	5.52	-0.48	0.63
16	1.00	$.8f_{16}^2 + 2f_{16}$	$.4f_{16}^2$	$.7u_{16}^2 + 3u_{16}$	25	20.68	-2.14	0.00
17	.98	$.5f_{17}^2 + 3f_{17}$	$.5f_{17}^2$	$2u_{17}^2 + u_{17}$	14	16.15	2.15	9.59
18	1.00	$.7f_{18}^2 + f_{18}$	$.7f_{18}^{2}$	$u_{18}^2 + u_{18}$	0	0.00	0.00	1.00
19	1.00	$.6f_{19}^2 + 4f_{19}$	$.4f_{19}^{2}$ $.5f_{20}^{2}$	$u_{19}^2 + 2u_{19}$	16	14.58	-1.00	0.00
20	.98	$1.1f_{20}^2 + 5f_{20}$	$.5f_{20}^2$	$.8u_{20}^2 + u_{20}$	13	7.34	-0.62	0.00

Example 5 Results (cont'd)

As expected, most of the computed capacity changes were negative as a result of the diminished demand for blood at our demand points.

The projected demand values were as follows:

$$v_1^* = 5.52$$
, $v_2^* = 35.25$, and $v_3^* = 23.02$.

The value of the total cost criterion for this Example was 51, 221.32.

Summary and Conclusions

we developed a sustainable supply chain network design model for a highly perishable health care product – that of human blood. Our model:

- captures the perishability of this life-saving product through the use of arc multipliers;
- contains discarding costs associated with waste/disposal;
- determines the optimal enhancement/reduction of capacities as well as the determination of the capacities from scratch;
- can capture the cost-related effects of shutting down specific modules of the supply chain due to an economic crisis;
- handles uncertainty associated with demand points;
- assesses shortage/surplus penalties at the demand points, and
- quantifies the supply-side risk associated with procurement.

Thank You!



Home About Background Activities Publications Media Links What's New Se



The Virtual Center for Supernetworks at the Isenberg School of Management, under the directorship of Anna Nagurney, the John F. Smith Memorial Professor, is an interdisciplinary center, and includes the Supernetworks Laboratory for Computation and Visualization.

Mission: The mission of the Virtual Center for Supernetworks is to foster the study and application of supernetworks and to serve as a resource to academia, industry, and government on networks ranging from transportation, supply chains, telecommunication, and electric power networks to economic, environmental, financial, knowledge and social

The Applications of Supernetworks Include: complex networks and decisionmaking: critical infrastructure from transportation to electric power and the Internet: financial. economic, and social networks; energy and the environment; global supply chain management; corporate social responsibility: risk management; network vulnerability, resiliency, and performance metrics; ecological networks; humanitarian logistics and healthcare.

















You are visitor number 75 264

to the Virtual Center for Supernetworks **GBY** Connecting



Google' Google Search