

Worst Case Analysis of Greedy,
Max-Regret and Other Heuristics for
Multidimensional Assignment and
Traveling Salesman Problems

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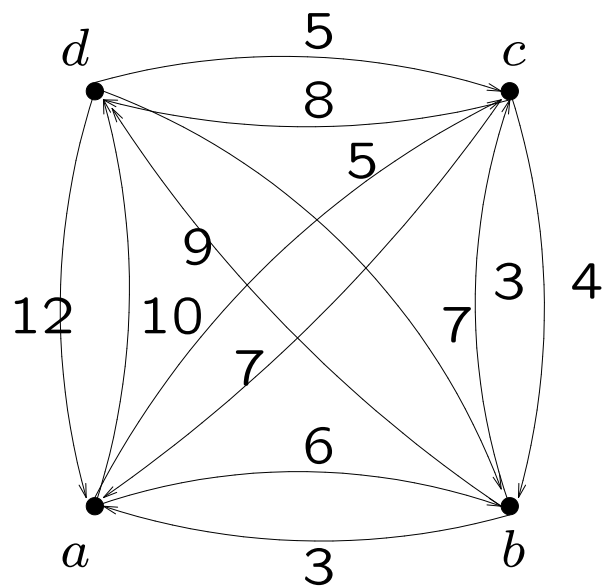
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Talk Overview

- ATSP
- Multidimensional Assignment Problem
- General results and results for other problems

ATSP



- A Hamilton cycle is a *tour*; $(n - 1)!$ tours
- Many heuristics (how to compare them?)
- Computational experiments
- Approximation and domination analyzes

Domination Number

- Introduced in 1997 by Glover and Punnen
- First results obtained by Rublineckii in 1973 [in Russian]
- For an ATSP heuristic H and ATSP instance I , $\text{domn}(H, I)$ is the number of tours in I of weight at least the weight of the tour obtained by H for I
- $\text{domn}(H, n) = \min\{\text{domn}(H, I) : |I| = n\}$

Use of Domination Number

- domn is invariant under linear transformations of arc weights
- Worst case comparison:
 $\text{domn}(H, n) > \text{domn}(H', n)$ for n large enough
Ben-Arieh, Gutin, Penn, Yeo and Zverovitch (2003) for Generalized ATSP
- Neighborhoods for local search: why some exponential-size neighborhoods perform bad while small-size ones perform well (Orlin et al.)?

Low Dom Number Heuristics

- Greedy and NN are of dom number 1 (may produce the unique worst possible tour), Gutin, Yeo and Zverovitch (2002)
- Any greedy-type heuristic is of dom number 1 (greedy-type introduced by Gutin, Vainshtein and Yeo, 2002, proved by Bendall and Margot, to appear)
- Max-Regret is of dom number 1 (Proc. WAOA'06)

Large Dom Number Heuristics

Any heuristic that always produces a tour of weight not worse than the average weight is of dom number at least $(n - 2)!$, $n \neq 6$ (n odd: Sarvanov, 1976, n even: Gutin and Yeo, 2002). Such heuristics are:

(a) Greedy-expectation algorithm (Gutin and Yeo, 2002)

(b) Vertex insertion algorithms (Lifshitz, 1973, Punnen and Kabadi, 2002)

(c) 3-Opt, Lin-Kernighan in polynomial time (Punnen, Margot and Kabadi, 2003)

In computational experiments low dom number algorithms often perform worse than large dom number algorithms

ATSP-Max-Regret

ATSP-Max-Regret-FC: Set $W = T = \emptyset$. While $V \neq W$ do the following: For each $i \in V \setminus W$, compute two lightest arcs (i, j) and (i, k) that are feasible additions to T , and compute the difference $\Delta_i = |w_{ij} - w_{ik}|$. For $i \in V \setminus W$ with maximum Δ_i choose the lightest arc (i, j) , which is a feasible addition to T and add (i, j) to M and i to W .

ATSP-Max-Regret: The same with not only outgoing, but also incoming arcs considered.

Question (Ghosh et al., 2007): What is the dom number of ATSP-Max-Regret? (ATSP-Max-Regret often outperforms Greedy)

Dom Number of ATSP-Max-Regret

Theorem *Both ATSP-Max-Regret-FC and ATSP-Max-Regret are of dom number 1.*

Proof: Consider an instance of ATSP with vertex set $\{1, 2, \dots, n\}$, $n \geq 2$. The weights: $w_{ik} = \min\{i - k, 0\}$ for each $1 \leq i \neq k \leq n$, $i \neq n$, and $w_{nk} = -k$ for each $1 \leq k \leq n - 1$.

Modify the weights: $w'_{ij} = w_{ij}$ unless $j = i + 1$ modulo n . Set $w'_{i,i+1} = -1 - \frac{1}{n+1}$ for $1 \leq i \leq n$.

ATSP-Max-Regret-FC will use w' .

Proof cont'd (a)

ATSP-Max-Regret-FC constructs the tour

$$T_{MR} = (1, 2, \dots, n, 1)$$

by choosing the arc $(n-1, n)$, the arc $(n-2, n-1)$, etc. The last two arcs are $(1, 2)$ and $(n, 1)$ (they must be included in the tour).

Indeed, initially $\Delta_{n-1} = \frac{n+2}{n+1} > \Delta_i = 1$ for each $i \neq n-1$. Once $(n-1, n)$ is added to T_{MR} , $\Delta_{n-2} = \frac{n+2}{n+1}$ becomes maximal, etc.

Let T', T'' be tours. Since $\sum_{(i,j) \in K_n^*} |w_{ij} - w'_{ij}| < 1$, $w(T') < w(T'')$ implies $w'(T') < w'(T'')$. Thus, in the rest of the proof we may use w rather than w' .

Proof cont'd (b)

Observe $w(T_{MR}) = -n$.

Let $T = (i_1, i_2, \dots, i_n, i_1)$ be an arbitrary tour, where $i_1 = 1$. Let $i_s = n$, $P = (i_1, i_2, \dots, i_s)$.

$$w(P) = \sum_{k=1}^{s-1} \min\{0, i_k - i_{k+1}\} \leq$$

$$\sum_{k=1}^{s-1} (i_k - i_{k+1}) = i_1 - i_s.$$

Thus, $w(P) \leq 1 - n$ and $w(P) = 1 - n$ iff $i_1 < i_2 < \dots < i_s$.

Since $i_s = n$, the weight of the arc (i_s, i_{s+1}) equals $-i_{s+1}$. Thus, $w(T) \leq 1 - n - i_{s+1}$ and $w(T) \geq w(T_{MR})$ iff $i_{s+1} = 1$ and $i_1 < i_2 < \dots < i_s$. We conclude that $w(T) \geq w(T_{MR})$ iff $T = T_{MR}$. \square

Multidimensional Assignment Problem (s -AP)

Let $X = \{1, 2, \dots, n\}^s$. Each vector $e \in X$ is assigned a real weight $w(e)$.

For a vector e , e_j denotes its j th coordinate. Vectors e, f are *independent* if $e_j \neq f_j$ for each $j = 1, 2, \dots, s$.

An *assignment* is a set of n independent vectors. The *weight* of an assignment $\{e^1, \dots, e^n\}$ is $\sum_{i=1}^n w(e^i)$.

The aim: to find an assignment of minimum weight.

Applications: In tracking objects, e.g., 5-AP to track elementary particles at CERN

Max-Regret by Balas and Saltzman (1991)

We have a partial assignment A .

Choose $j \in \{1, \dots, s\}$ and $m \in \{1, \dots, n\}$.

Choose two lightest vectors e, f with $e_j = f_j = m$ and independent from vectors in A . Find the regret $\Delta_{j,m} = |w(e) - w(f)|$.

Choose j, m with maximum $\Delta_{j,m}$ and the lightest e with $e_j = m$. Add e to A .

Low Dom Number Heuristics

In comput. experiments Balas and Saltzman (1991) show Max-Regret outperforms Greedy. In comput. experiments Robertson (2001) shows the heuristics are of similar performance.

Theorem For s -AP, $s \geq 3$, both Greedy and Max-Regret are of dom number 1.

Theorem For 2-AP, Greedy is of dom number 1.

Theorem For 2-AP, Max-Regret is of dom number at most 2^{n-1} .

Large Dom Number Heuristics

Theorem For s -AP, if a heuristic H always produces an assignment of weight at most the average weight \bar{w} of an assignment, then we have $\text{domn}(H, n) \geq ((n-1)!)^{s-1}$.

Proof: Consider an instance \mathcal{I} of s -AP. C denotes the set of all vectors of \mathcal{I} with the first coordinate equal 1; $\mathcal{P} = \{A_f : f^1 \in C\}$, where $A_f = \{f^1, f^2, \dots, f^n\}$ is an assignment with $f_j^i = f_j^1 + i - 1$ (modulo n), $j = 1, 2, \dots, s$.

Each vector is in exactly one A_f and, thus, \mathcal{P} is a partition of $X = X_1 \times X_2 \times \dots \times X_s$, $X_i = \{1, 2, \dots, n\}$, into assignments.

Since $\sum_{f \in C} w(A_f) = w(X)$, $|C| = n^{s-1}$ and $\bar{w} = w(X)/n^{s-1}$, the heaviest assignment A_h in \mathcal{P} is of weight at least \bar{w} .

Proof cont'd

Let $S(X_i)$ be the set of all permutations on X_i ($2 \leq i \leq s$) and let $\pi_2 \in S(X_2), \dots, \pi_s \in S(X_s)$. To obtain $\mathcal{P}(\pi_2, \dots, \pi_s)$ from \mathcal{P} , replace f_j^i with $\pi_j(f_j^i)$ for each $j \geq 2$ and $i = 1, 2, \dots, n$. Thus, we obtain a family

$$\mathcal{F} = \{\mathcal{P}(\pi_2, \dots, \pi_s) : \pi_2 \in S(X_2), \dots, \pi_s \in S(X_s)\}$$

of partitions of X into assignments. The family consists of $(n!)^{s-1}$ partitions. We may choose the heaviest assignment in each partition and, thus, obtain a family \mathcal{A} of assignments of weight at least \bar{w} .

It's possible to prove that no assignment in \mathcal{A} can be repeated more than n^{s-1} times. Since \mathcal{A} has $(n!)^{s-1}$ assignments with repetitions, we can find (in \mathcal{A}) $((n-1)!)^{s-1}$ distinct assignments of weight at least \bar{w} . \square

ROM Heuristic

Recursive Opt Matching (ROM):

Theorem For 3-AP, 3-Opt (by Balas and Saltzman) is of dom number at least $((n - 1)!)^2$.

Compute a new weight $\bar{w}(i, j) = w(X_{ij})/n^{s-2}$, where X_{ij} is the set of all vectors with last two coordinates equal i and j , respectively. Solving the 2-AP with the new weights, find an optimal assignment $M = \{(i, \pi_s(i)) : i = 1, 2, \dots, n\}$, where π_s is a permutation.

Continue for coordinates $s - 2$ and $s - 1$ with M 'fixing' coordinates $s - 1$ and s , etc. Use $X' = \{1, 2, \dots, n\}^{s-1}$ instead of X .

Theorem

$\text{domn}(ROM, n) \geq ((n - 1)!)^{s-1}$.

Proof

It suffices to show that the assignment obtained by ROM is of weight at most $\bar{w} = w(X)/n^{s-1}$, the average weight of an assignment. By induction on $s \geq 2$. Clearly the assertion holds for $s = 2$ and consider $s \geq 3$. Observe that

$$\bar{w} = \frac{1}{n} \sum_{i=1}^n \sum_{j=1}^n \bar{w}(i, j) \geq \sum_{i=1}^n \bar{w}(i, \pi_s(i)) = \frac{w'(X')}{n^{s-2}}.$$

Let $A = \{(g^1, \pi_s(1)), \dots, (g^n, \pi_s(n))\}$ be an assignment obtained by ROM, where $g^i \in X'$ such that $g_{s-1}^i = i$ for every $i = 1, \dots, n$. Let $A' = \{g^1, \dots, g^n\}$. Then by induction hypothesis, $\bar{w}' = w'(X')/n^{s-2} \geq w'(A') = w(A)$ and we are done. \square

Preliminary Computational Experiments (by Gerold Jäger)

100 random examples with $n=8$, $s=6$, entries in $[0,1000]$ (sum of all solutions):

1. Greedy: 58310
2. Recursive Opt Matching: 53820
3. Balas-Saltzman: 54637
4. Shifted Recursive Opt Matching: 36878

Results for Other Problems

1. Independence Systems (IS). Sufficient conditions for Greedy to be of domination number 1, Gutin and Yeo, 2002
2. IS. Weights form a finite set. Necessary and sufficient conditions for Greedy to be of domination number 1, Bang-Jensen, Gutin and Yeo, 2004.
3. IS. Sufficient conditions for greedy-type heuristics to be of domination number 1, Bendall and Margot, to appear
4. Dom analysis of various problems: Alon, Gutin and Krivelevich (2004), Berend, Skiena and Twitto (submitted), Gutin, Vainshtein and Yeo 2003, Gutin, Jensen and Yeo (2006), etc.