## SMOOTHING DIGITIZED CONTOURS

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Abstract. We give a fast linear-time algorithm for finding a smooth polygonal approximation to a digitized contour such that the digitization of the polygonal contour matches the original input. The polygonal contour has the minimum possible number of inflections and obeys a localized best-fit property. Most of the vertices lie on a grid whose resolution is only twice that of the pixel grid, and the algorithm can be modified to force all vertices to obey this restriction.

Key Words. smoothing, digitized image, polygonal outline, polygonization

1. Introduction. In graphics and computer typesetting, black and white images are commonly represented as arrays of pixels, and it is convenient to manipulate such images via the contours that describe black-white boundaries. For instance, if pixels are thought of as unit squares that tile the plane, the contours that describe the digitized image of a letter "R" might appear as in Figure 1a. Of course, the smoother contours shown in Figure 1b are a much better rendition of the original design on which the figures are based. In fact, the digitized image was created by applying a digitization process to smooth contours similar to those in Figure 1b. We give a smoothing algorithm that computes Figure 1b from Figure 1a and guarantees that the digitization process would regenerate Figure 1a from Figure 1b. This inverse digitization property is particularly important because it enables the algorithm to eliminate digitization noise without obliterating subtle curves.

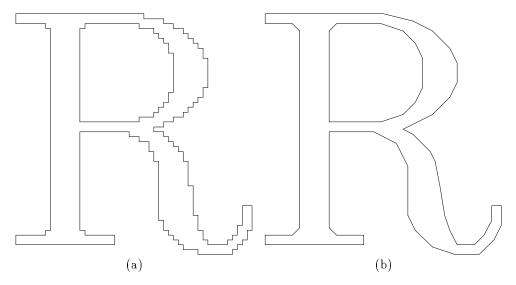


Fig. 1. (a) Digitized contours for a letter "R" from Knuth's Computer Modern Roman typeface [4]; (b) corresponding smooth polygonal contours.

The digitization of a smooth polygonal contour  $C_p$  is a well defined curve that can be computed by a fast linear algorithm similar to Bresenham's algorithm [1]. Applying the function

$$\rho(x,y) = \left( \lceil x - \tfrac{1}{2} \rceil, \lfloor y + \tfrac{1}{2} \rfloor \right)$$

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to each point on  $C_p$  yields a list of points

$$(1) (m_0, n_0), (m_1, n_1), (m_2, n_2), \ldots, (m_k, n_k),$$

where  $(m_k, n_k) = (m_0, n_0)$ . Ordinarily, each difference  $(m_i, n_i) - (m_{i-1}, n_{i-1})$  will be either  $(0, \pm 1)$  or  $(\pm 1, 0)$  and we may define the digitization  $\mathcal{D}(C_p)$  to be the polygonal contour whose vertices are given by (1). The definition fails when a negatively sloped portion of  $C_p$  passes through ambiguous points of the form  $(m+\frac{1}{2}, n+\frac{1}{2})$  for integers m and n, but we can get by with non-ambiguous polygonal contours that contain no such points. (A definition that does handle ambiguous points appears in [3].) Note that the definition also applies to non-closed curves in which case  $(m_k, n_k) \neq (m_0, n_0)$ .

The digitization  $\mathcal{D}(C_p)$  is a polygonal contour made of integer-length vertical and horizontal edges as in Figure 1a. Any contour  $C_d$  with this property is called a *digital contour*. In order to have  $\mathcal{D}(C_p) = C_d$ , a contour  $C_p$  must pass sequentially through the squares

(2) 
$$S(m_0, n_0), S(m_1, n_1), S(m_2, n_2), \ldots, S(m_k, n_k),$$

where

$$S(m,n) = \{(x,y) \mid m - \frac{1}{2} < x \le m + \frac{1}{2} \text{ and } n - \frac{1}{2} \le y < n + \frac{1}{2} \}$$

is the set of all (x, y) such that  $\rho(x, y) = (m, n)$ . Our smoothing process generates a polygonal contour  $C_p$  with this property. There are many ways to do this, but this paper gives a new approach that yields polygonal contours with many desirable properties.

The algorithm has important applications to image processing and computer type-setting. It does an excellent job of converting bitmap fonts into polygonal outlines, and it performs well in typesetting applications where the vertices of the polygonal outlines must be given to a limited precision. Other possible applications include scaling bitmap fonts and processing black and white images. The image processing application uses thresholding and edge detection to produce digital contours, while subsequent analysis requires smooth contours. (See Montanari [5] for a more detailed discussion.)

Previous work has concentrated on minimizing the total length of the smoothed contour subject to constraints on the deviation from the digitized input. In [8], Sklansky deals with constraints essentially identical to those implied by the inverse digitization property, and he outlines a fast linear algorithm for computing a polygonal contour that will be convex if any convex contour satisfies his constraints. In [5], Montanari gives a slower algorithm that applies to any digitized contour and deals with broader classes of constraints.

By minimizing total length, Sklansky and Montanari obtain polygonal contours that tend to maximize local deviations from the digitized input. A better approach is to reduce local deviations as much as possible without introducing any extraneous points of inflection. We therefore begin with two theorems that provide insight into the set of all contours that minimize the number of points of inflection without violating the inverse digitization property.

2. Polygonization and Single-Quadrant Paths. We now restrict our attention to portions of digital contours that are monotone in x and y. If  $\sigma_1$  and  $\sigma_2$  are two of the four direction vectors  $(0,\pm 1)$  and  $(\pm 1,0)$  where  $\sigma_1 \neq \pm \sigma_2$ , a  $\{\sigma_1,\sigma_2\}$  digital

path is a polygonal line joining the integer grid points

$$(m_0, n_0), (m_1, n_1), (m_2, n_2), \ldots, (m_k, n_k),$$

in order, where each difference  $(m_i, n_i) - (m_{i-1}, n_{i-1})$  is of the form  $l_i \sigma_1$  or  $l_i \sigma_2$  for some positive integer  $l_i$ .

In general, a path that is a  $\{\sigma_1, \sigma_2\}$  digital path for some directions  $\sigma_1$  and  $\sigma_2$  is called a *single-quadrant digital path*. Since the smoothing process converts single-quadrant digital paths into more general *single-quadrant polygonal paths*, we define a  $\{\sigma_1, \sigma_2\}$  polygonal path to be a polygonal line joining points

$$(3) (x_0, y_0), (x_1, y_1), (x_2, y_2), \ldots, (x_k, y_k),$$

in order, where each difference  $(x_i, y_i) - (x_{i-1}, y_{i-1})$  is  $\alpha_{1i}\sigma_1 + \alpha_{2i}\sigma_2$  for some  $\alpha_{1i}, \alpha_{2i} \ge 0$ . In other words,  $(x_i, y_i) - (x_{i-1}, y_{i-1})$  is a nonnegative combination of  $\sigma_1$  and  $\sigma_2$ .

A  $\{\sigma_1, \sigma_2\}$  digital path with vertices  $(m_i, n_i)$  for  $0 \le i \le k$  is said to be polygonizable if for each positive i < k, at least one of the vectors  $(m_i, n_i) - (m_{i-1}, n_{i-1})$  or  $(m_{i+1}, n_{i+1}) - (m_i, n_i)$  has length one. The polygonization of such a path is the  $\{\sigma_1, \sigma_2\}$  polygonal path obtained by connecting with straight lines the points  $\frac{1}{2}(m_i+m_{i+1}, n_i+n_{i+1})$  for  $0 \le i < k$ . In other words, the polygonization  $\mathcal{P}(P)$  of a single-quadrant digital path P is obtained by connecting the midpoint of every edge in P as shown in Figure 2.

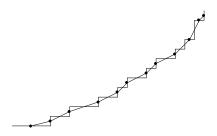


Fig. 2. A polygonizable  $\{(1,0), (0,1)\}$  digital path and its polygonization. The bold dots mark the points  $\frac{1}{2}(m_i+m_{i+1}, n_i+n_{i+1})$  that are vertices of the polygonization.

If a  $\{\sigma_1, \sigma_2\}$  polygonal path P passes through points  $P_1$  and  $P_2$  in that order,  $P_2 - P_1$  will always be a nonnegative combination of  $\sigma_1$  and  $\sigma_2$ . Thus the *subpath* of P from  $P_1$  to  $P_2$  contains all points z on P such that both  $z - P_1$  and  $P_2 - z$  are nonnegative combinations of  $\sigma_1$  and  $\sigma_2$ . Another important operation on single-quadrant polygonal paths is a *simple extension* formed by lengthening the first and last edges; i.e., Q is a simple extension of P if P is a subpath of Q, whose first and last edges overlap nontrivially with those of Q.

The smoothing algorithm assumes that polygonal paths can be extended in order to achieve a desired digitization. Thus we say that a single-quadrant polygonal path P is digitally extensible to a single-quadrant digital path Q if Q can be expressed as the digitization of a subpath of a simple extension of P.

The two theorems below on digital extensibility are based on Lemmas 5.2.1 and 5.2.2 of [3], so we only justify them informally here. Theorem 2.2 depends on the concept of a digital transformation; i.e., an affine transformation that is 1 to 1 and onto on the set  $\mathbb{Z}^2 + (\frac{1}{2}, \frac{1}{2})$  of points (x, y) where  $x \equiv y \equiv \frac{1}{2}$  modulo 1. All such transformations are of the form  $T(x, y) = (x - \frac{1}{2}, y - \frac{1}{2})A + (m + \frac{1}{2}, n + \frac{1}{2})$ , where m and n are integers and A is a two-by-two integer matrix with determinant  $\pm 1$ .

Some applications of digital transformations are discussed by Rothstein and Weiman in [7, 6]

Theorem 2.1. If P is a polygonizable single-quadrant digital path, then the polygonization  $\mathcal{P}(P)$  is digitally extensible to P.

Theorem 2.2. If P is a polygonizable single-quadrant digital path and there is a digital transformation T such that  $T(\mathcal{P}(P))$  is a  $\{(1,0),(0,1)\}$  digital path, then a non-ambiguous  $\{(1,0),(0,1)\}$  polygonal path Q is digitally extensible to  $T(\mathcal{P}(P))$  if and only if  $T^{-1}(Q)$  is digitally extensible to P.

Figure 3 illustrates the ideas behind Theorems 2.1 and 2.2. Figure 3a shows a  $\{(1,0),(0,1)\}$  digital path P and the corresponding sequence of squares (2) through which a path must pass in order to have P as its digitization. The polygonization  $\mathcal{P}(P)$  is not shown in the figure, but it could be formed by joining the bold dots with a polygonal line. Theorem 2.1 simply says that  $\mathcal{P}(P)$  passes through the indicated squares when its first and last edges are suitably extended.

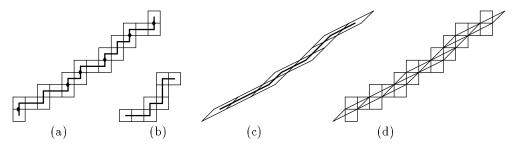


Fig. 3. (a) A  $\{(1,0), (0,1)\}$  digital path and squares where  $\rho(x,y)$  is constant; (b) the polygonization transformed to yield a  $\{(1,0), (0,1)\}$  digital path; (c) the same figure untransformed; (d) a superposition that illustrates Theorem 2.2.

If P is the path shown in Figure 3a, Figure 3b shows the  $\{(1,0), (0,1)\}$  digital path  $T(\mathcal{P}(P))$ , where

$$T(x,y) = (\tfrac{1}{2} + x - y, \, -x + 2y) \quad \text{and} \quad T^{-1}(x,y) = (2x + y - 1, \, x + y - \tfrac{1}{2}).$$

The squares in Figure 3b are where a path Q must pass in order to have  $T(\mathcal{P}(P))$  as its digitization. Figure 3c is simply the result of applying  $T^{-1}$  to Figure 3b to obtain a sequence of parallelograms through which  $T^{-1}(Q)$  must pass. Figure 3d illustrates Theorem 2.2 by superimposing the relevant portions of Figures 3a and 3c. The path  $T^{-1}(Q)$  that passes through the parallelograms when its first and last edges are extended will also pass through the squares in Figure 3d when the first and last edges are extended sufficiently.

3. The Smoothing Algorithm. Theorems 2.1 and 2.2 suggest a simple recursive smoothing algorithm: Given a digital contour, break it into single-quadrant digital paths, take the polygonization of each path, and repeat for each polygonization that can be transformed into a single-quadrant digital path. If P is a single-quadrant digital path, Theorem 2.1 ensures that  $\mathcal{P}(P)$  is digitally extensible to P. If there is a single-quadrant digital path of the form  $T(\mathcal{P}(P))$ , Theorem 2.2 ensures that we may recursively obtain a polygonal path Q that is digitally extensible to  $T(\mathcal{P}(P))$ , and  $T^{-1}(Q)$  will be digitally extensible to P. Naturally, we extend the concept of polygonization so that we can compute  $\mathcal{P}(\mathcal{P}(P))$  instead of  $T^{-1}(\mathcal{P}(T(\mathcal{P}(P))))$ . Thus, the recursive process just repeatedly polygonizes the polygonization.

To facilitate such repeated polygonization, we need a representation that captures the essential similarities between digital paths and digital transformations thereof. Any polygonal path with rational edge slopes can be represented as a root  $(x_0, y_0)$  and a list of edge nodes  $e_1, e_2, \ldots, e_k$ , where each edge node  $e_i$  contains a numeric length parameter  $l(e_i)$  and a direction vector  $d(e_i)$  of the form  $(a_i, b_i)$  for relatively prime integers  $a_i$  and  $b_i$ . The corresponding polygonal path has vertices  $(x_i, y_i) = \sum_{j=1}^{i} l(e_j) \cdot d(e_j)$ .

If  $\sigma_1 \neq \pm \sigma_2$  are two vectors of the form  $(\pm 1, 0)$  or  $(0, \pm 1)$ , a  $\{\sigma_1, \sigma_2\}$  digital path representation has  $(x_0, y_0) \in \mathbf{Z}^2$  and each  $l(e_i) \in \mathbf{Z}$ , where  $\mathbf{Z}$  is the set of integers. In addition, the directions  $d(e_1), d(e_2), \ldots, d(e_k)$  alternate  $\sigma_1, \sigma_2, \sigma_1, \sigma_2$ , etc. A digital transformation of such path has a very similar representation with  $\sigma_1$  and  $\sigma_2$  replaced by their images  $(a_1, b_1)$  and  $(a_2, b_2)$ . The root will satisfy

(4) 
$$x_0 \equiv \frac{1}{2}(1 + a_1 + a_2) \pmod{1}$$
 and  $y_0 \equiv \frac{1}{2}(1 + b_1 + b_2) \pmod{1}$ ,

and the directions will alternate between  $(a_1, b_1)$  and  $(a_2, b_2)$ , where  $a_1b_2 - a_2b_1 = \pm 1$ . If  $a_1a_2 + b_1b_2 > 0$ , the path is a generalized digital path, or more specifically, an  $\{(a_1, a_2), (b_1, b_2)\}$  digital path. If no two adjacent  $l(e_1)$  and  $l(e_{i+1})$  are both greater than one, the path is polygonizable and the polygonization may be computed as shown in Figure 4.

```
\begin{split} & \textbf{function} \ polygonize(x_0, y_0, e_1, e_2, \dots, e_k); \\ & i \leftarrow 1; \ n \leftarrow 0; \\ & (\bar{x}_0, \bar{y}_0) \leftarrow (x_0, y_0) + \frac{1}{2}l(e_1) \cdot d(e_1); \\ & \textbf{while} \ i < k \\ & \textbf{do} \ \{ \ j \leftarrow i+1; \\ & \textbf{while} \ j < k \ \text{and} \ l(e_{j-1}) = l(e_{j+1}) \ \textbf{do} \ j \leftarrow j+1; \\ & n \leftarrow n+1; \\ & d(\bar{e}_n) \leftarrow l(e_i) \cdot d(e_i) + l(e_{i+1}) \cdot d(e_{i+1}); \\ & l(\bar{e}_n) \leftarrow (j-i)/2; \\ & i \leftarrow j; \ \} \\ & \textbf{return} \ (\bar{x}_0, \bar{y}_0, \bar{e}_1, \bar{e}_2, \dots, \bar{e}_n); \end{split}
```

Fig. 4. A function that computes the polygonization.

Instead of giving up when polygonize is not directly applicable to its own output, we can try to break the offending path into subpaths that are polygonizable generalized digital paths. The break routine in Figure 5 shows how to find the required break-points, and the polygonize\_ext routine polygonizes the resulting subpaths and copes with non-integral  $l(e_1)$  and  $l(e_k)$ . These routines use a function  $\tau(e_i, e_j) = a_i b_j - a_j b_i$ , where  $d(e_i) = (a_i, b_i)$  and  $d(e_j) = (a_j, b_j)$  for edges  $e_i$  and  $e_j$ ; the sign of  $\tau(e_i, e_{i+1})$  determines whether the path turns left or right between  $e_i$  and  $e_{i+1}$ . Note that the argument to break is assumed to be part of a contour so that we may examine the edges  $e_0$  and  $e_{k+1}$  before and after the path to be broken up.

Applying break to the polygonization in Figure 6a, successive values of  $\tau(e_i, e_{i+1})$  are 2, -1, 1, 1, -1, 1, -1, 1, 1 - 1, and we obtain the six subpaths labeled A through F in Figure 6b. Subpaths A, C and E are not polygonizable because they each contain only one edge; subpath B is a polygonizable  $\{(2,1), (3,1)\}$  digital path; subpath D is a polygonizable  $\{(2,1), (1,1)\}$  digital path when polygonize\_ext extends the last edge  $e_{11}$  so that  $l(e_{11})$  is 1 as shown in Figure 6c instead of  $\frac{1}{2}$  as shown in Figure 6b.

```
function polygonize\_ext(x_0, y_0, e_1, e_2, \dots, e_k);
r \leftarrow l(e_1);
l(e_1) \leftarrow \lceil (e_1) \rceil;
l(e_k) \leftarrow \lceil (e_k) \rceil;
if l(e_k) \neq r then (x_0, y_0) \leftarrow (x_0, y_0) + (r - l(e_1)) \cdot d(e_1);
return polygonize(\bar{x}_0, \bar{y}_0, \bar{e}_1, \bar{e}_2, \dots, \bar{e}_n);

procedure break(e_0, e_1, e_2, \dots, e_k, e_{k+1});
for i \leftarrow 1, 2, \dots, k
do if \tau(e_{i-1}, e_i) and \tau(e_i, e_{i+1}) are both positive or both negative then break before and after e_i;
for i \leftarrow 1, 2, \dots, k - 1
do if |\tau(e_i, e_{i+1})| > 1 or l(e_i) > 1 and l(e_{i+1}) > 1
then break between e_i and e_{i+1};
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Fig. 5. Routines for breaking a polygonal path into pieces that are digital transformations of single-quadrant polygonal paths.

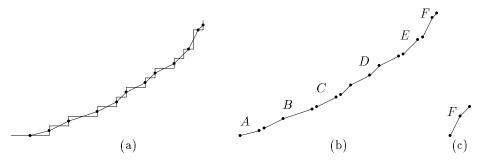


Fig. 6. (a) A single-quadrant digital path and its polygonization with the vertices of the polygonization marked by dots; (b) subpaths produced by break; (c) a subpath produced by polygonize\_ext.

In general, the effect of the first loop in the break routine is to ensure that the subpaths generated will contain only inflection edges. The effect of the second loop is to ensure that  $\tau(e_i, e_{i+1})$  alternates between 1 and -1. Since the input to break is either a digital contour or the polygonization of a polygonizable generalized digital path, it can be shown that the direction vectors come from a restricted set where  $\tau(e_{i-1}, e_i) = -\tau(e_i, e_{i+1})$  implies  $d(e_{i-1}) = d(e_{i+1})$ . Lemma 3.2 from Section 3.1 then shows that the subpaths produced by break are legal input to  $polygonize\_ext$ : We can construct digital transformations that map them into simple extensions of single-quadrant digital paths.

After repeatedly calling break and  $polygonize\_ext$ , we obtain a list of polygonal paths with gaps where edges from polygonize need to be extended. The complete smoothing algorithm shown in Figure 7 uses a routine  $join\_paths$  to eliminate the gaps by extending the first and last edges of each subpath until they intersect. For brevity, we use P, Q, and R in lieu of edge list representations for polygonal paths and lists of polygonal paths; we also introduce a routine pop(P) that extracts the edges before the first breakpoint in P. When P is a complete contour consisting of sequences of edge nodes with breakpoints in between, pop(P) removes one such sequence of edge nodes.

```
\begin{aligned} & \textbf{function} \ smooth(\textbf{P}); \\ & \textbf{Make} \ Q' \ \text{empty}; \\ & Q \leftarrow break(P); \\ & \textbf{while} \ Q \ \text{is not empty} \\ & \left\{ \ R \leftarrow pop(Q); \\ & \textbf{if} \ R \ \text{contains more than one edge} \\ & \textbf{then append} \ break(polygonize\_ext(R)) \ \text{to the beginning of} \ Q \\ & \textbf{else append} \ R \ \text{to the end of} \ Q'; \\ & \left\} \\ & \textbf{return} \ join\_paths(Q'); \end{aligned}
```

Fig. 7. The smoothing algorithm.

Figure 8 illustrates the action of smooth on a simple convex shape. When we first reach the top of the **while** loop, the digital contour is broken into subpaths as shown by the dots at subpath boundaries in Figure 8a. We next reach the top of the loop after polyginizing the first subpath and breaking it up as shown in Figure 8b. The **if** test fails in the next two iterations, but the following iteration does call polygonize, obtaining a single edge of slope  $\frac{3}{2}$  as shown in Figure 8c. The **while** loop terminates after 27 more iterations, obtaining Figure 8d.

The action of  $join\_paths$  on Figure 8d is fairly simple: The list of polygonal paths can be thought of as a polygonal contour with gaps, and  $join\_paths$  eliminates each gap by extending the surrounding edges until they intersect. For example,  $join\_paths(Q)$  may be implemented as shown in Figure 9 if Q is a circular doubly linked list of edge and gap nodes where each gap node  $g_i$  contains a vector  $\delta(g_i) = (x_i, y_i)$ . (We use  $\otimes$  for the operation  $(x_1, y_1) \otimes (x_2, y_2) = x_1y_2 - x_2y_1$  so that  $\tau(e_i, e_j) = d(e_i) \otimes d(e_j) = -d(e_j) \otimes d(e_i)$  for all i and j.)

The join\_paths function given in Figure 9b contains a loop that eliminates edge nodes with nonpositive length parameters. Such "backward edges" are not very common, but they can be created when the loop "for each gap  $g \in Q$  do  $remove\_gap(g)$ " is applied to a Q that contains edges and gaps like those in Figure 10a and Table 1a.

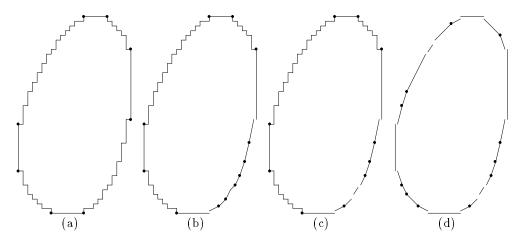


Fig. 8. Snapshots of Q and Q' in smooth (P) after (a) zero iterations, (b) one iteration, (c) four iterations, (d) 31 iterations.

```
function join\_paths(Q);
                                                            for each gap g \in Q do remove\_gap(g);
procedure remove\_gap(g);
                                                            repeat
Let e_1 and e_2 be the edges surrounding og; e \leftarrow true;
q \leftarrow \tau(e_1, e_2);
l(e_1) \leftarrow l(e_1) - (d(e_2) \otimes \delta(g)) / q;
l(e_2) \leftarrow l(e_2) + (d(e_1) \otimes \delta(g)) / q;
l(e_2) \leftarrow l(e_3) + (d(e_3) \otimes \delta(g)) / q;
                                                                for each edge e \in Q such that l(e) \leq 0
                                                                do \{ z \leftarrow l(e) \cdot d(e); 
                                                                          Replace e with a gap g and set \delta(g) \leftarrow z;
Remove g from the linked list;
                                                                          remove\_gap(g);
                                                                          \textit{done} \leftarrow \textit{false}; \}
                                                            \mathbf{until}\ done\ ;
                                                             {\bf return}\ Q;
                                 (a)
                                                                                                   (b)
```

Fig. 9. Routines for eliminating gaps by extending the surrounding edges.

Parts b, c, and d of Figure 10 and Table 1 illustrate subsequent steps in the execution of  $join\_paths(Q)$ : The backward edge is replaced by a gap that is then removed.



Fig. 10. Polygonal paths (a) as given to join\_paths, (b) after remove\_gap(g) for each gap g, (c) after removing the backward edge, and (d) as computed by join\_paths.

$\operatorname{node}$	l	d	$\delta$				
$\overline{e_1}$	1	(3, -1)		_	node	l	d
${g}_1$			$(0,-\frac{1}{2})$		$e_1$	$1\frac{1}{2}$	(3, -1)
$e_2$	1	(1, 0)	. 2		$e_2$	-2	(1, 0)
$g_2$			$(0,\frac{1}{2})$		$e_3$	$1\frac{1}{2}$	(3, 1)
$e_3$	1	(3, 1)		-			
(a)				(b)			
		(4)				(~)	
		(4)				(5)	
		(4)				(5)	
node	l	d	δ		node	<i>i</i>	d
node	$l = 1\frac{1}{2}$				node	<i>l</i>	$\frac{d}{(3-1)}$
	_	$\frac{d}{(3,-1)}$	$\delta = (-2,0)$		$\overline{e_1}$	$\begin{array}{ c c } & l \\ \hline & 1\frac{1}{6} \end{array}$	(3, -1)
$e_1$	$\frac{l}{1\frac{1}{2}}$ $1\frac{1}{2}$	d				<i>l</i>	
$egin{array}{c} e_1 \ g_3 \end{array}$	_	$\frac{d}{(3,-1)}$			$\overline{e_1}$	$\begin{array}{ c c } & l \\ \hline & 1\frac{1}{6} \end{array}$	(3, -1)

Edge and gap data corresponding to Figures 10a-d.

The motivation for the use of backward edges in  $join\_paths$  depends on the concept of a  $polygonal\ tracing$  as described by Guibas, Ramshaw, and Stolfi [2]. Using the extension of digitization to polygonal tracings as described in [3],  $join\_paths$  is intended to keep the digitization of Q invariant. A proof would be too tedious to be of much value, but extensive practical experience indicates that the smoothing algorithm given in this section does transform any digital contour P into a polygonal path whose digitization is P. Theorems 2.1 and 2.2 clearly show the validity of repeated polygonization, but it is harder to show that  $join\_paths$  always yields the desired results.

3.1. The Localized Best-Fit Property. If the smoothing algorithm computes a polygonal contour Q from a digital contour P, the localized best fit property is that local changes to Q can only increase the maximum deviation

$$\max(\max_{z\in P}\operatorname{dist}(z,Q),\ \max_{z\in Q}\operatorname{dist}(z,P))$$

between P and Q, where dist() refers to Euclidean distance. The local changes that we shall consider are shifts of a single edge parallel to itself as shown in Figure 11. In Figure 11a, the maximum deviation of the digital path shown in bold from the slope 1 edge of the polygonization is  $1/\sqrt{8}$  pixels. Shifting the slope 1 edge as shown in Figure 11b or 11c doubles the maximum deviation.

A more precise statement of the best-fit property is "If Q = smooth(P) has an edge node e with  $l(e) \ge 1$ , then a parallel shift of the edge represented by e cannot

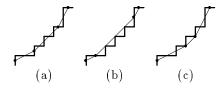


Fig. 11. A digital path with its polygonization (a), and perturbations (b) and (c).

decrease the maximum deviation between P and Q in the neighborhood of e." In other words, all sufficiently long edges of Q are optimally placed, where the definition of "sufficiently long" is liberal enough to apply to a large majority of the edges in Q. Theorem 3.3 below captures the key idea behind the best-fit property while avoiding the messy details of how the edges of Q fit together. Using the idea that for purposes of measuring maximum deviation, any sufficiently long rational-slope line segment behaves like an infinite line, we define a line  $\ell$  of rational slope to be optimally placed if it is non-ambiguous and the maximum deviation between  $\ell$  and its digitization  $\mathcal{D}(\ell)$  is less than the maximum deviation between  $\mathcal{D}(\ell)$  and any other line  $\ell'$  parallel to  $\ell$ .

LEMMA 3.1. A line of the form qx - py = c for relatively prime integers p and q is optimally placed if and only if  $c \equiv \frac{1}{2}(1 + p + q)$  (modulo 1).

*Proof.* Let  $\ell$  be the line qx - py = c and let  $\mathcal{D}(\ell)$  be its digitization. The line  $\ell$  is ambiguous if qx - py = c has solutions where  $x \equiv y \equiv \frac{1}{2} \pmod{1}$ . In this case  $c \equiv \frac{1}{2}p + \frac{1}{2}q$  and  $\ell$  is not optimally placed.

The remaining case is where  $\ell$  is non-ambiguous and  $c \not\equiv \frac{1}{2}p + \frac{1}{2}q$ . The maximum deviation between  $\ell$  and  $\mathcal{D}(\ell)$  is  $\max_{(x,y)\in V} \alpha |qx - py - c|$ , where  $\alpha = 1/\sqrt{p^2 + q^2}$  and  $V = \{ \rho(x,y) \mid qx - py = c \}$  is the set of vertices of  $\mathcal{D}(\ell)$ . Since  $\ell$  is non-ambiguous,

$$V = \{ (x, y) \in \mathbf{Z}^2 \mid qx' - py' = c \text{ for some } (x', y') \text{ where } \rho(x', y') = (x, y) \}$$

$$= \{ (x, y) \in \mathbf{Z}^2 \mid qx' - py' = c \text{ for some } (x', y') \text{ where } x' \not\equiv \frac{1}{2} \not\equiv y' \text{ and } \rho(x', y') = (x, y) \}$$

$$= \{ (x, y) \in \mathbf{Z}^2 \mid qx' - py' = c \text{ for some } (x', y') \text{ where } |x' - x| < \frac{1}{2} \text{ and } |y' - y| < \frac{1}{2} \}$$

$$= \{ (x, y) \in \mathbf{Z}^2 \mid |qx - py - c| < \frac{1}{2} |p| + \frac{1}{2} |q| \},$$

where **Z** is the set of integers. Since  $\{qx - py \mid (x,y) \in \mathbf{Z}^2\} = \mathbf{Z}$ , the deviation between  $\ell$  and  $\mathcal{D}(\ell)$  is

$$\max_{(x,y)\in V} \alpha |qx - py - c| = \max_{n\in N} \alpha |n - c|$$

where

$$\begin{split} N &= \{\, q \, x - p y \mid (x,y) \in V \,\} \\ &= \{\, n \in \mathbf{Z} \mid |n-c| < \tfrac{1}{2} \, |p| + \tfrac{1}{2} \, |q| \,\}. \end{split}$$

If  $\ell'$  is a line qx - py = c', the above analysis shows that the maximum deviation E(c') between  $\ell'$  and  $\mathcal{D}(\ell)$  is  $\max_{n \in N} \alpha |n - c'|$ . Clearly,  $\ell$  is optimally placed if and only if the minimum of E(c') occurs at c' = c. Since  $c \not\equiv \frac{1}{2}p + \frac{1}{2}q$  (modulo 1), we have  $c + r \not\in \mathbf{Z}$  where  $r = \frac{1}{2}|p| + \frac{1}{2}|q|$ . Thus the largest and smallest elements of N are  $\lfloor c + r \rfloor$  and  $1 + \lfloor c - r \rfloor$ , and the minimum of E(c') occurs at

$$\frac{1 + \lfloor c - r \rfloor + \lfloor c + r \rfloor}{2}.$$

Setting E(c') = c yields  $c = \frac{1}{2}(1 + \lfloor c - r \rfloor + \lfloor c + r \rfloor) = \frac{1}{2}(1 + \lfloor c - r \rfloor + \lfloor c - r \rfloor + \lfloor c - r \rfloor + 2r) = \frac{1}{2}(1 + 2r) + \lfloor c - r \rfloor \equiv \frac{1}{2}(1 + 2r) \pmod{1}$ . Thus  $c \equiv \frac{1}{2}(1 + |p| + |q|) \equiv \frac{1}{2}(1 + p + q)$  as desired.  $\square$ 

Lemma 3.2. Digital transformations map optimally placed lines into optimally placed lines.

*Proof.* A digital transformation  $T(x,y)=(x-\frac{1}{2},y-\frac{1}{2})A+(m+\frac{1}{2},n+\frac{1}{2})$  maps the line qx-py=c into a line q'x-p'y=c' where

$$c' = T(x,y)(q',-p')^T = (x,y)A(q',-p')^T - (\tfrac{1}{2},\tfrac{1}{2})A(q',-p')^T + (m+\tfrac{1}{2},n+\tfrac{1}{2})(q',-p')^T$$

when qx-py=c. Thus we may set  $A(q',-p')^T=(q,-p)^T$  and  $c'=c-(\frac{1}{2},\frac{1}{2})A(q',-p')^T+(m+\frac{1}{2},n+\frac{1}{2})(q',-p')^T$ . When  $c\equiv\frac{1}{2}(1+p+q)$  (modulo 1), we have

$$\begin{split} c' &\equiv c - (\frac{1}{2}, \frac{1}{2})(q, -p)^T + (m + \frac{1}{2}, n + \frac{1}{2})(q', -p')^T \\ &= c - \frac{1}{2}q + \frac{1}{2}p + \frac{1}{2}q' - \frac{1}{2}p' + (m, n)(q', -p')^T \\ &\equiv \frac{1}{2}(1 + p + q) + \frac{1}{2}p + \frac{1}{2}q - \frac{1}{2}p' - \frac{1}{2}q' \\ &\equiv \frac{1}{2}(1 + p' + q'). \end{split}$$

Theorem 3.3. If  $\ell$  is the line determined by some edge in a polygonal contour computed by the smoothing algorithm, then  $\ell$  is optimally placed.

*Proof.* Let us say that an edge is well placed when the line determined by that edge is. We show by induction that all edges computed by the algorithm are well placed. Clearly, all edges the initial digital contour are well placed because we may apply Lemma 3.1 with  $c \in \mathbf{Z}$  and  $(p,q) \in \{(0,\pm 1), (\pm 1,0)\}$ .

All other edges are created by polygonizing to a digital transformation of a polygonizable single-quadrant digital path. It suffices to show that the line joining the midpoints of two adjacent edges of such a path is optimally placed. Then Lemma 3.2 allows us to apply the digital transformation that makes the edges in question go from  $(-l_1,0)$  to (0,0) and from (0,0) to  $(0,l_2)$ , where  $l_1$  and  $l_2$  are positive integers not both greater than one. By Lemma 3.1, the line  $l_2x - l_1y = -\frac{1}{2}l_1l_2$  joining the midpoints is optimally placed because

$$-\frac{1}{2}l_1l_2 = \frac{1}{2}(1+l_1+l_2) - \frac{1}{2}(1+l_1)(1+l_2) \equiv \frac{1}{2}(1+l_1+l_2) \pmod{1}.$$

Thus Lemma 3.2 shows that all new edges are well placed, and the theorem follows.  $\Pi$ 

**3.2.** Minimization of Inflections. An important property of smoothing algorithms is that they should eliminate extraneous points of inflection. Figure 1 exemplifies this: The digital contours in Figure 1a have 118 inflections while the smoothed versions in Figure 1b have only eight. In fact, no polygonal contours whose digitizations match Figure 1a can have fewer than eight inflections. The purpose of this section is to show why the results of *smooth* are always optimal in this way.

Since inflections in polygonal contours occur at edges, it is convenient to refer to edges as either inflection edges or plain edges. The main idea is that the break procedure only inserts gap nodes next to plain edges or at "sharp bends" where any contour with the correct digitization must have similar bends. The algorithm proceeds until edge and gap nodes alternate and each inflection edge is surrounded by either "sharp bends" or sequences of plain edges.

When counting the inflection edges in a polygonal path P with edges  $e_1$ ,  $e_2$ , ...,  $e_k$ , we need to know the signs of  $\tau_{\text{in}} = \tau(e_0, e_1)$  and  $\tau_{\text{out}} = \tau(e_k, e_{k+1})$ , where  $e_0$  and  $e_{k+1}$  are the edges adjacent to P in the contour. Let  $\mathcal{I}(\tau_{\text{in}}, P, \tau_{\text{out}})$  be the number of i where  $1 \leq i \leq k$  and  $\tau(e_{i-1}, e_i)$  and  $\tau(e_i, e_{i+1})$  have opposite signs.

Let P be a path such that some simple extension P' of P is a single-quadrant digital path. Given the signs of  $\tau_{\rm in}$  and  $\tau_{\rm out}$ , we may apply smooth to yield a polygonal path  $\mathcal{S}(\tau_{\rm in}, P, \tau_{\rm out})$ . The computation is equivalent to evaluating  $break(polygonize\_ext(P))$  and then recursively applying smooth to each resulting subpath. We show by induction on the number of levels of recursion that

$$\mathcal{I}(\tau_{\text{in}}, \mathcal{S}(\tau_{\text{in}}, P, \tau_{\text{out}}), \tau_{\text{out}}) \leq \mathcal{I}(\tau_{\text{in}}, R, \tau_{\text{out}})$$

for any path R that is digitally extensible to P'.

Given a polygonizable  $\{(1,0), (0,1)\}$  digital path P with polygonization  $Q = \mathcal{P}(P)$ , break(Q) produces subpaths  $Q_1, Q_2, \ldots, Q_n$  of Q, where each  $Q_i$  ends at the starting point of  $Q_{i+1}$ . For each  $Q_i$  there is a subpath  $P_i$  of P such that  $Q_i = \mathcal{P}(P_i)$ . The induction step depends on using the following lemma to apply Theorem 2.2 to each  $P_i$  and  $Q_i$ .

LEMMA 3.4. If  $P_1, P_2, \ldots, P_n$  are as defined above, then any polygonal path R that is digitally extensible to P is the union of disjoint subpaths  $R_1, R_2, \ldots, R_n$  such that each  $R_i$  is digitally extensible to  $P_i$ .

Proof. Each interior vertex of  $\mathcal{P}(P)$  lies at the midpoint of some edge  $e_j$  of P where  $l(e_{j-1}) \neq l(e_{j+1})$ . We cannot have  $l(e_j) > 1$ , because the polygonizability of P would force  $l(e_{j-1}) = l(e_{j+1}) = 1$ . The perpendicular bisector of  $e_j$  will be the boundary between the the squares  $S(m_{j-1}, n_{j-1})$  and  $S(m_j, n_j)$  from (2), where  $(m_{j-1}, n_{j-1})$  and  $(m_j, n_j)$  are the endpoints of  $e_j$ . Any path whose digitization is P must cross this boundary exactly once.

Since R is digitally extensible to P, there is a subpath R' of a simple extension of R such that P is the digitization of R'. Breaking R' where it crosses the perpendicular bisectors of the  $e_j$  that contain endpoints of  $Q_1, Q_2, \ldots, Q_n$ , we obtain disjoint subpaths  $R'_1, R'_2, \ldots, R'_n$  such that each  $R'_i$  is digitally extensible to  $P_i$ . To find  $R_1, R_2, \ldots, R_n$  as required by the lemma, we modify  $R'_1, R'_2, \ldots, R'_n$  so that their union is R instead of R': Any initial subpath of R not in R' is appended to  $R'_1$ ; any left-over final subpath of R is appended to  $R'_n$ ; any  $R_i$  not fully contained in R has its first and/or last edges shortened as necessary.  $\square$ 

Each path  $Q_i$  produced by break(Q) is either a line segment or is extensible to a polygonizable generalized digital path. If I is the set of all i such that  $Q_i$  is nontrivial, then for each  $i \in I$ , there is a digital transformation  $T_i$  such that a simple extension of  $T_i(Q_i)$  is a polygonizable  $\{(1,0),(0,1)\}$  digital path. Lemma 3.4 produces an  $R_i$  that is digitally extensible to  $P_i$ , and Lemma 3.5 below allows us to assume without loss of generality that  $T_i(R_i)$  is a  $\{(1,0),(0,1)\}$  polygonal path for each  $i \in I$ . Thus Theorem 2.2 shows that for  $i \in I$ ,  $T_i(R_i)$  is digitally extensible to a simple extension of  $T_i(Q_i)$ .

Lemma 3.5. For any polygonal path R with subpaths  $R_1, R_2, \ldots, R_n$  satisfying Lemma 3.4, there exists a path R' that satisfies the lemma and has the following properties: (1) There must be subpaths  $R'_1, R'_2, \ldots, R'_n$  of R such that the transformation  $T_i$  that maps  $Q_i$  into a  $\{(1,0), (0,1)\}$  digital path maps  $R'_i$  into a  $\{(1,0), (0,1)\}$  polygonal path. (2) All nonzero  $\tau_{in}$  and  $\tau_{out}$  satisfy  $\mathcal{I}(\tau_{in}, R', \tau_{out}) \leq \mathcal{I}(\tau_{in}, R, \tau_{out})$ .

*Proof.* A complete proof would be rather long and involved but the basic idea is fairly simple: For each i,  $R'_i$  is formed by applying a series of local modifications

to  $R_i$ , where each modification preserves the digitization of  $R_i$  and does not increase the number of inflection edges. For example, if  $R_i$  passes through the squares in Figure 3d then  $R_i'$  will pass through the parallelograms in that figure. The process of constructing  $R_i'$  from i is similar to removing from  $R_i$  all portions not contained in the parallelograms and then replacing the missing subpaths with portions of the parallelogram boundaries.  $\square$ 

Since  $T_i(Q_i)$  requires one less level of polygonization than P, the induction hypothesis shows that if  $T_i(R_i)$  is a  $\{(1,0),(0,1)\}$  digital path then

(5) 
$$\mathcal{I}(\tau_i, T_i(Q_i), \tau_i') \le \mathcal{I}(\tau_i, T_i(R_i), \tau_i')$$

for any nonzero  $\tau_i$  and  $\tau_i'$ . Since affine transformations preserve inflections, (5) still holds when  $T_i(Q_i)$  and  $T_i(R_i)$  are replaced by  $Q_i$  and  $R_i$ .

For each two integers m < n belonging to I such that  $i \notin I$  for m < i < n, let  $\tau'_m = \tau_n = \tau(\bar{e}'_m, \bar{e}_n)$ , where  $\bar{e}'_m$  is the last edge in  $Q_m$  and  $\bar{e}_n$  is the first edge in  $Q_n$ . A simple argument based on Lemma 3.5 allows us to assume that  $\tau(\hat{e}'_m, \hat{e}_n)$  and  $\tau(\bar{e}'_m, \bar{e}_n)$  have the same sign, where  $\hat{e}'_m$  is the last edge in  $R_m$  and  $\hat{e}_n$  is the first edge in  $R_n$ . Thus even if R has no inflection edges except those in some nontrivial  $R_i$ , each  $R_i$  will contain as many inflection edges as the corresponding  $Q_i$ . Hence the total number of inflection edges in R will be at least as great as the total for Q.

- 4. Refinements. Figures 4, 5, 7, and 9 give a complete implementation of the smoothing algorithm based on lists of edge and gap nodes, and we have already seen that the smoothing algorithm produces a polygonal path whose digitization matches its input while obeying a localized best-fit property and minimizing the number of inflections. These properties do not leave much flexibility, but we can still make some refinements that tend to simplify the smoothed contours.
- **4.1.** Adjustments Prior to Polygonization. The first refinement affects the way  $polygonize\_ext$  prepares a path for polygonization. After extending the first and last edges so that  $l(e_1)$  and  $l(e_k)$  are integers, it is sometimes desirable to make further adjustments as shown in Figure 12. In each part of the figure, bold dots denote points where gap nodes are inserted into the data structure to delimit subpaths suitable for polygonization. Figure 12a shows a  $\{(1,0),(0,1)\}$  digital path where  $l(e_1)$  is less than  $l(e_3)$ ,  $l(e_5)$ , and  $l(e_7)$ . As shown in Figure 12b,  $e_1$  may be replaced by a gap node because the smoothed version of  $e_2$ ,  $e_3$ , ...,  $e_7$  will begin with an edge that can be extended so that its digitization matches  $e_1$ . Figures 12c and 12d show a  $\{(1,0),(0,1)\}$  digital path where an extra gap may be inserted to force the  $e_1$  edge to be treated as a separate subpath.

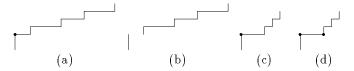


Fig. 12. Digital paths before and after adjustments that help prepare for polygonization.

Figures 13a and 13b show that the effect of the adjustment shown in Figures 12a and 12b is to eliminate an unnecessary edge of slope  $\frac{1}{2}$  between the vertical edge and the edge of slope  $\frac{2}{7}$ . As Figures 13c and 13d demonstrate, the effect of the adjustment of Figures 12c and 12d is to replace an edge of slope  $\frac{1}{3}$  with an edge of slope 0. The

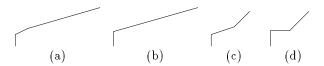


Fig. 13. The results of applying smooth to Figures 12a-12d.

slope 0 edge is an inflection edge, but this is safe here because an inflection is required in any case.

The routines in Figure 14 illustrate one way to decide when to apply the above adjustments. A call to *choice* with arguments  $l(e_1)$ ,  $l(e_2)$ , ...,  $l(e_k)$  returns *omit* if  $e_1$  should be removed as in Figure 12b; it returns *separate* if  $e_1$  should be a separate subpath as in Figure 12d; and it returns *use* if no adjustment is desirable. A call to *choice* with arguments  $l(e_k)$ ,  $l(e_{k-1})$ , ...,  $l(e_1)$  makes a similar decision for the last edge  $e_k$  of a polygonizable generalized digital path with edge list  $e_1, e_2, \ldots, e_k$ .

```
function tread\_change(l_1, l_2, \ldots function choice(l_1, l_2, \ldots, l_k); for i \leftarrow 2, 4, 6, \ldots if l_2 > 1 or k = 2 then return use; do \{ if l_i > 1 then return -1; if l_3 < l_1 or k = 3 if i = k then return 0; then return (if l_3 < l_1 - 1 then separate else use) if l_{i+1} > l_{i-1} then return l_i if l_3 > l_1 or tread\_change(l_3, l_4, \ldots, l_k) > 0 if i + 1 = k then return 0; then return omit if l_{i+1} < l_{i-1} then return -1; else return use; (b)
```

Fig. 14. Functions for deciding on adjustments prior to polygonization.

4.2. Retaining Long Inflection Edges. Figure 15a shows one of the polygonizable single-quadrant digital subpaths produced by applying break to Figure 1a. Figure 15b shows that its polygonization has one inflection edge and the slope of that edge is -12. Treating the long edge separately by inserting gap nodes at the dots in Figure 15c yields Figure 15d after polygonizing and Figure 15e after using join\_paths. Figures 15b and 15e both have the same number of inflections and they both satisfy the localized best-fit property, but Figure 15e may be preferred for aesthetic reasons.

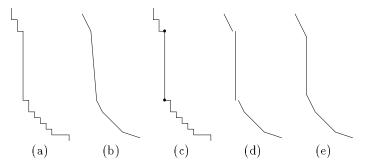


Fig. 15. A single-quadrant digital path and smoothed versions computed with and without retaining the longest inflection edge.

Given a polygonizable generalized digital path P with edge list  $e_1, e_2, \ldots, e_k$ , an edge  $e_i$  is a long inflection edge if any path digitally extensible to P must have an

inflection edge in the neighborhood of  $e_i$ . More precisely,  $l(e_i)-1$  must be greater than both  $l(e_{i-2})$  and  $l(e_{i+2})$ , and we must have i-2>1 and i+2< k. Aesthetic qualities govern the choice of which long inflection edges to retain, but a good compromise to to retain those where  $l(e_i) > l(e_{i-2}) + l(e_{i+2})$ . This may be done by adding the following loop to the end of the *break* procedure in Figure 5:

for 
$$i \leftarrow 4, 5, \dots, k-3$$
  
do if  $l(e_i) > l(e_{i-2}) + l(e_{i+2})$  then break before and after  $e_i$ .

**4.3.** Redirecting Free Edges. When smooth is applied to the contour P shown in Figure 16a, the first thing it does is call break(P), and this immediately breaks the contour into eight trivial subpaths. This means that smooth(P) just returns the original contour P. For the bitmap font application, the contour shown in Figure 16b can be a much preferable result, and this is obtained by a simple refinement to the smoothing algorithm.



Fig. 16. (a) A digital contour; (b) a smoothed version produced by the refinement discussed in Section 4.3.

When break(P) makes the short vertical edges in Figure 16a into independent subpaths, they introduce some flexibility into the smoothing algorithm. Ordinarily, if Q is a digital path containing only one edge and Q' is a single-edge polygonal path that is digitally extensible to Q, the possibilities for the direction of Q' form a narrow range centered on the direction of Q, but the range becomes  $180^{\circ}$  when the Q edge is only one unit long. When the Q edge is also an inflection edge, we call it a free edge. We can recognize free edges by using the following block in place of the statement "append R to the end of Q'" in the smooth function:

```
{ Let e be the edge in R;

if l(e) = 1 and e is an inflection edge then mark e as a free edge;

Append R to the end of Q' }.
```

Free edges need to be processed by a separate loop after all other edge directions have been determined. We therefore insert

for each edge 
$$e \in Q'$$
 do if  $adjustable(e)$  then  $adjust(e)$ 

just before the **return** statement in *smooth*, where *adjustable* and *adjust* are the routines shown in Figure 17.

5. Conclusion. The basic smoothing algorithm is detailed in Figures 4, 5, 7, and 9 in Section 3. In spite of the complexity of the ideas behind the algorithm, the actual code is fairly simple. It is a fast linear-time algorithm that generates contours that have the localized best-fit property and minimize inflections subject to the constraint that the original contour must be the digitization of the smoothed contour. These properties do not uniquely determine the result of the smoothing algorithm, but the algorithm is intended to produce what might be regarded as the "best and simplest" polygonal approximation.

```
function adjustable(e);
if e is not a free edge then return false;
Let e_{-} and e_{+} be the edge nodes before and after e_{:}
Let g_{-} and g_{+} be the gap nodes before and after e;
return d(e_{-}) = d(e_{+}) and |\tau(e_{-}, e)| = 1 and \delta(g_{-}) = \delta(g_{+}) = (0, 0);
procedure adjust(e);
Let g_{-} and g_{+} be the gaps before and after e;
Let e_{-} and e_{+} be the edges before and after e_{:}
Let g_{-2} and g_{+2} be the gaps before e_{-} and after e_{+};
Let e_{-2} and e_{+2} be the edges before e_{-} and after e_{+};
if adjustable(e_{-2}) then l_{-} \leftarrow \frac{1}{2}l(e_{-})
else { remove\_gap(g_{-2}); l_{-} \leftarrow l(e_{-}); } }
if adjustable(e_{+2}) then l_{+} \leftarrow \frac{1}{2}l(e_{+})
else { remove\_gap(g_{+2}); l_{+} \leftarrow \overline{l}(e_{+}); } }
\delta(g_-) \leftarrow \delta(g_-) - \min(l_-, l_+) \cdot d(e_-);
\delta(g_+) \leftarrow \delta(g_+) - \min(l_-, l_+) \cdot d(e_-);
d(e) \leftarrow d(e) + 2\min(l_-, l_+) \cdot d(e_-);
```

Fig. 17. Routines for adjusting free edges.

It is difficult to define exactly what is meant by "best and simplest," but the refinements discussed in Section 4 give a rough idea of the range of possibilities. The implementations make reasonable choices as to how and when each refinement should take effect, but the choices could be changed depending on the definition of "best and simplest."

Another possible change is related to the claim made in the abstract that "most of the vertices lie on a grid whose resolution is twice that of the pixel grid." For instance, Figure 1b contains 55 vertices, all but one of which lie in  $(\frac{1}{2}\mathbf{Z})^2$ , where  $\frac{1}{2}\mathbf{Z}$  is the set of integer multiples of  $\frac{1}{2}$ . All the vertices do lie on such a grid if  $join\_paths$  exits with  $l(e) \in \frac{1}{2}\mathbf{Z}$  for all edges e, but Table 1 shows that this is not always true. The exceptions tend to occur where relatively long edges with nonsimple slopes intersect, and the exceptional vertices can usually be repositioned slightly without doing much damage to the smoothed contour. (It sometimes helps to apply a digital transformation before rounding the vertices to grid points.)

Another claim that we have not addressed so far is the linear running time of the smoothing algorithm. The intuition behind this is that the number of edges in the polygonization of a path P is at most a fixed fraction of the number of edges in P so that the total number of edges created during polygonization can be found by summing a geometric series. A simple inductive argument shows that the sum of  $l(e_i)$  over all edges  $e_i$  created during polygonization is no more than the total number of inflections removed by the algorithm. For example, Figure 1a has 136 edges and 118 inflections, while Figure 1b has only eight inflections. Thus 110 inflections were removed and the 33 edges created by polygonize had  $\sum_i l(e_i) = 48$ . The linear bound on  $\sum_i l(e_i)$  implies a linear bound on the number of edge nodes created, and this also bounds the number of gap nodes created.

The edge and gap implementation given here can be made reasonably efficient, but it should be noted that this model was chosen for ease of understanding rather than for maximum efficiency. In practice, most of the execution time is spent in the initial phases of the algorithm when there are a few single-quadrant digital subpaths,

each containing a large number of edges. Rather than storing such subpaths as lists of edge nodes, we can save time and space by taking advantage of the fact that within a generalized digital path  $e_1, e_2, \ldots, e_k$ , the direction  $d(e_i)$  depends only on whether the index i is odd or even. Thus all  $l(e_i)$  values can be stored in a single array L, and a  $\{d_0, d_1\}$  digital path with edges  $e_1, e_2, \ldots, e_k$  becomes a 4-tuple  $(i_0, k, d_0, d_1)$  that refers to  $L[i_0+1], L[i_0+2], \ldots, L[i_0+k]$ , where  $d(e_i) = d_{i \mod 2}$  and  $l(e_i) = L[i_0+i]$ . This way, only the 4-tuples need to be stored in a doubly linked list and the gap displacements  $\delta(g_i)$  can be merged with the 4-tuples. The only drawback is that it is not possible to represent an arbitrary sequence of edges without inserting undesired subpath boundaries. This means that the polygonize and break routines have to be merged together because the output of polygonize might not be a generalized digital path.

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