Antarctic Ice Sheet Balance Velocities from Merged Point and Vector Data

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Abstract

We estimate Antarctic ice-flow balance-velocities, which are the average speeds that ice must flow downslope through a volume assuming that there are equal amounts of ice entering and leaving the ice volume. We use the OSU Digital Elevation Model (DEM) of Antarctica, recent ice accumulation rate data, and the BEDMAP ice thickness data compilations for Antarctica to characterize the physical properties of the ice sheet that are included in the balance velocity calculation. We adapt a flux algorithm from the hydrology literature that enables us to calculate the flux distribution from any cell in any order. Flux from one cell to its neighbors is partitioned as a simple function of surface slope direction. Digitized flow stripe directions from satellite images minimize errors in flow direction where surface slopes are low or complex. We estimate errors in balance velocity arising from errors in the data and show semi-quantitatively how properties of the algorithm bias the balance velocity result. We find a favorable comparison between our model and observed velocity data as well as the balance velocity patterns reported by other researchers.

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1 Introduction

A basic question about any ice sheet concerns whether or not it is thickening or thinning. One way to answer that question is to model the steady state properties of the ice sheet and then compare the model with measurements. A common approach along this line infers the steady-state depth-averaged velocities from a model based on mass continuity. The continuity calculation is initiated with measurements of ice sheet surface elevation, ice thickness, and surface and basal accumulation rate. The calculated velocities are known as balance velocities because they represent the ice sheet depth-averaged speeds in the downslope direction that would result if the amount of snow added annually to a volume of the ice sheet equaled the amount of mass lost from that volume through advection or melting (Budd and others, 1971). If the surface balance velocities inferred from the (depth averaged) balance velocities are higher than the measured velocities, the ice sheet is thickening and vice versa.

In this paper, we present computed surface balance velocities and error budgets for the Antarctic Ice Sheet. Our analysis builds on previous research (Budd and others, 1971; Budd and others, 1982; Budd and Smith, 1985; Budd and Warner, 1996; Bamber and others, 2000a, 2000b; Huybrechts and others, 2000) by adapting and modifying an algorithmic approach developed in the hydrology community (Costa-Cabral and Burges, 1994). Our numerical approach, which casts the outflow from a grid cell into a downstream flux-matrix, refines the balance velocity calculation by incorporating vector data on flow direction from the Radarsat Antarctic Mapping Project image mosaic (Jezek,

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1999). The numerical implementation of our algorithm allows us to calculate a total error budget that includes errors arising from the input data and suggests how our algorithm may bias the final result. We validate our computed surface balance velocities and velocity errors through comparison with measured velocities on profiles encompassing the Lambert Glacier and across ice streams draining into the Filchner Ice Shelf.

2. Balance velocity equation

The mathematical expression of balance velocity for continuous fields is given by Budd and Warner (1996), who also discuss approaches for discretizing the model. On a regular grid, the total flux from any grid cell is treated as a scalar with a fractional part of the flux going in orthogonal grid (*x* and *y*) directions. The portion of the flux in the *x* direction $(|F_x|)$ is related to an equivalent balance velocity component $(|V_x|)$ as:

$$\left|F_{x}\right| = \left|V_{x}\right| \cdot H \cdot W = \left|V\right| \cdot \left|\cos\theta\right| \cdot H \cdot W \tag{1}$$

where *H* is the ice thickness, *W* is the square-grid cell-size dimension, |V| is the magnitude of the balance velocity vector and θ is the flow direction with respect to the x-axis. Similarly:

$$\left|F_{y}\right| = \left|V_{y}\right| \cdot H \cdot W = \left|V\right| \cdot \left|\sin\theta\right| \cdot H \cdot W$$
(2)

and the total flux (|F|) for a cell is:

$$\left|F\right| = \left|F_{x}\right| + \left|F_{y}\right| \tag{3}$$

Equations (1), (2) and (3) yield:

$$\left|V\right| = \frac{\left|F\right|}{H \cdot W \cdot \left(\left|\sin\theta\right| + \left|\cos\theta\right|\right)} \tag{4}$$

For comparison with measured surface data, Equation (4) is modified by the ratio (r) between balance velocity and the ice surface balance velocity ($|V_s|$), which takes into account the fact that the velocity decreases from the surface towards the bed. The value of r ranges from 0.8 to 1.0 (Paterson, 1994) and we choose 0.9. $|V_s|$ then becomes

$$\left|V_{s}\right| = \frac{\left|F\right|}{H \cdot W \cdot \left(\left|\sin\theta\right| + \left|\cos\theta\right|\right) \cdot r}$$
(5)

The value of r depends on properties such as ice temperature and basal conditions, so r will vary across the ice sheet. We select a constant value reflecting our lack of knowledge about spatial variability in r.

3. Numerical Evaluation of Balance Velocity

Several approaches for evaluating balance velocities start by determining the positions of flow lines. Balance velocities are calculated either by integrating fluxes over areas bounded by flow lines (Budd and others, 1971; Joughin and others, 1997) or by integrating point measurements along a flow line (Radok and others, 1982). Although the technique accurately captures balance velocity physics, flow line determination over long distances is prone to small errors in the surface slope and can be sensitive to the order (upstream versus downstream) in which flow lines are evaluated (Radok and others, 1982). Most recent automatic approaches seem to over-concentrate flux towards the center of convergent flow (Bamber and others, 2000b).

Budd and Smith (1985) devised an automatic gridded technique that does not rely on an initial estimate of flow line positions. Budd and Warner (1996) and Fricker and others (2000) provide quantitative descriptions of a flow-line independent gridding technique that includes four generic steps: Digital Elevation Model (DEM) preparation; flow direction estimation; flux estimation; and final balance velocity estimation. The first and last steps are common between different investigators (Budd and Warner, 1996; Bamber and others, 2000a, b) and we essentially adopt a similar approach as discussed below. There are different options for the second and third steps in which we have made modifications to procedures for estimating flow direction and flux. We discuss these below in comparison with the work of other investigators.

3.1 DEM preparation

We use the 1 km DEM from Liu and others (1999) in Arc/Info grid format. We filter the DEM using a running, locally-adaptive, Gaussian weighting window corresponding to an averaging dimension of 20 times the ice thickness (Paterson, 1994; Bamber and others, 2000b). At the margin of Antarctica, the DEM is smoothed to the local mean to avoid edge effects.

Residual sinks (that is a local depression from the mean slope, resulting from data measurement error or data rounding) are filled using the Arc/Info grid function. The function raises the elevation of the sink to match the lowest height of the eight neighboring cells.

The filtered DEM is bilinearly resampled into a 20km cell-size grid, which corresponds to the cell size used by Budd and Warner (1996). The bilinear interpolation function provided by Arc/Info 8.1 assigns the value for an output cell using the weighted average of the four closest input cell centers to the output cell center. Ice flow directions are determined from terrain slopes calculated from the 20 km data.

3.2 Flow direction estimation

Flow direction is calculated by estimating the direction of steepest slope using a DEM. Budd and Warner (1996) calculate the slope components (α_x and α_y) for cell $x_{i,j}$ having elevation $E_{i,j}$ using the elevations of four cardinal nearest-neighbors surfaceelevations $E_{i-1,j}$, $E_{i,j+1}$, $E_{i+1,j}$ and $E_{i,j-1}$, where subscripts *i* and *j* represent the row and column indices of the DEM grid:

$$\alpha_{x} = (E_{i+1,j} - E_{i-1,j})/2W$$
(6)

$$\alpha_{y} = (E_{i,j-1} - E_{i,j+1})/2W \tag{7}$$

where W is the grid-cell dimension. The flow direction (θ) is given by

$$\sin\theta = \frac{\alpha_y}{\alpha}, \ \cos\theta = \frac{\alpha_x}{\alpha}$$
 (8)

where α is the magnitude of the slope.

We use the Costa-Cabral and Burges (1994) fitting plane algorithm to derive flow direction from the DEM. The algorithm uses the four most widely separated cell

(diagonal cells about the evaluation cell) heights $(E_{i-1,j-1}, E_{i-1,j+1}, E_{i+1,j-1})$ to fit a plane. The algorithm assigns the facing direction of the plane to the flow direction. The slope components (α_x, α_y) for cell $x_{i,j}$ are:

$$\alpha_x = \frac{E_{i-1,j-1} - E_{i-1,j+1} + E_{i+1,j-1} - E_{i+1,j+1}}{2W}$$
(9)

$$\alpha_{y} = \frac{E_{i+1,j+1} - E_{i-1,j+1} + E_{i+1,j-1} - E_{i-1,j-1}}{2W}$$
(10)

The flow direction (θ) is given by

$$\theta = \tan^{-1}(\frac{\alpha_y}{\alpha_x}) \tag{11}$$

The flow directions in either low-slope (Lea, 1992; Tarboton, 1997; Liang and Machay, 1997 and 2000) or highly convergent regimes are unreliable. To mitigate this problem we incorporate flow-direction information from imagery. In regions where the ice has been, or at least is assumed to be, in equilibrium, available information on flow stripes from the Radarsat Antarctic Mapping Project (RAMP) imagery is taken to accurately reflect the flow-direction. Our approach is to use the two end-points of each segment in the flow-stripe-vector data-set to calculate the flow-direction for that segment. We then convert the flow-stripe vector-data into grid-data, giving the grid-cell value the flowdirection of the closest line segment to the cell center if the line segment is within the cell. We merge the flow-stripe directions with surface-slope-derived flow-directions by preferentially selecting the flow stripe orientation. Incorporation of vector data mitigates the sensitivity of flow direction to DEM errors in low slope areas and to some extent, regions of converging flow. Flow stripes can incorrectly bias the result if the flow field has evolved over time. This is probably a greater problem on ice shelves, which are not treated in this analysis.

3.3 Flux and velocity estimation

Given estimates of flow direction and mass discharge per cell, partition schemes are used to model how much mass flows from one cell into one or two neighboring cells. Budd and Warner (1996) partition the mass into downstream cells by multiplying the total scalar flux out of a grid cell by terms defined as

$$\lambda_m = \frac{|\sin\theta|}{|\sin\theta| + |\cos\theta|}$$
(12)

.

$$\lambda_n = \frac{|\cos\theta|}{|\sin\theta| + |\cos\theta|} \tag{13}$$

Here m and n refer to the orthogonal grid directions.

As noted by Budd and Warner, the actual choice of partitioning scheme has little effect on the calculation (though as noted later the choice can play a role in the error model).

We favor the Costa-Cabral and Burges (1994) partition scheme wherein partitioning is proportional to areas defined by segmenting the cell into two regions separated by the flow direction vector. The partitioning using the Costa-Cabral and Burges (1994) approach goes as:

$$\lambda_m = \frac{\tan\theta}{2} \qquad (14)$$

$$\lambda_n = 1 - \frac{\tan \theta}{2} \qquad (15)$$

Here the definition of θ is slightly modified so that $0^{\circ} \le \theta \le 45^{\circ}$. Now, we essentially define θ as the angle of the balance velocity vector with respect to the cardinal direction (x or y), whichever is closer to the balance velocity direction.

Partition schemes model how one cell discharges mass into neighboring cells. The next step is to form the drainage system using a flux algorithm model describing the contributions from all the cells. Budd and Warner (1996) begin by sorting the DEM in descending order of elevation. Their algorithm depends on DEM sorting order, which makes it sensitive to DEM error.

In another approach that builds on the idea of using flow bands to calculate flux, Castal-Cabral and Burges (1994), trace upslope the flow lines passing through cells which then define the boundaries of contributing areas. Because the algorithm relies on flow directions to define contributing areas, flow direction errors at the early stage of the calculation can create large errors in contributing areas.

In a related approach, Costa-Cabral and Burges (1994) treat each cell as an initial condition wherein the flux entering the cell is simply the surface accumulation rate (advected fluxes are not included in the flux of the cell chosen for the start of the calculation). The discharge flux from the initial cell is allowed to flow downslope as an advective term until it encounters cells that are sinks or leaves the terrain. Once the discharge from every cell is calculated into a so-called influence matrix bounded laterally

by flowlines, the total mass passing through each cell is summed (both accumulation and advection terms). In this way, the total mass passing through a particular cell equals the total mass drained by every upstream cell.

Figure 1a illustrates the influence matrix idea. The gray region is the influence matrix of the cell A. Every cell in the gray region receives some amount of advected mass from cell A. The boundaries of the influence matrix are flow lines so that there is no diffusion of mass across the boundary. Figure 1b is an example of how the flux into cell B is determined by a subset of flow lines which bound the dark gray region of Figure 1b. Though the influence matrix determined by flow lines is conceptually very accurate, it is hard to write an efficient and accurate computer program.



Figure 1. The influence matrix associated with cell A (Figure 1a) where every gray cell receives some amount of mass from cell A. The amount of mass received by cell B (Figure 1b) is determined by the careful downslope tracing of flow lines as described in the original Costa-Cabral and Burges (1994) approach. The accuracy of the estimated flux to cell B is strongly influenced by the accuracy of the derived flow lines that bound the dark gray region in Figure 1b.

We modify the downslope algorithm (Costa-Cabral and Burges, 1994) by focusing on the influence matrix aspect of their model but abandoning explicit calculation of flow lines. This allows for a tractable code while retaining the advantage that it does not depend on DEM sorting order. As noted later, this modification does result in artificial mass diffusion through the grid.

The influence matrix represents the flux distribution for a cell using only the accumulated mass at that cell as a flux source. Hence every cell is associated with a downstream influence matrix. The algorithm calculates the influence matrix for every cell and sums up influence matrices to get the total flux distribution. The algorithm builds an influence matrix cell by cell by successively partitioning mass from one cell into its downslope cells according to the partitioning scheme proposed by Costa-Cabral and Burges. Partitioning proceeds until the edge of the terrain is reached. In Figure 2, flux from the originating cell A is partitioned into B and C and so on. The cells with the same gray tones are processed at the same time. We call these cells the 'frontline' and some care has to be taken when complex terrain may cause subsequent backflow into these cells (note that this problem is avoided in DEM sorting methods). Both the influence matrix and DEM sorting techniques suffer from diffusion, by which we mean the artificial distribution of mass away from boundaries that would properly be identified as flow lines (Tarboton, 1997).

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Figure 2. Modified DEMON-Downslope algorithm (influence matrix algorithm).

Since the influence matrix algorithm does not use the DEM order to process cells, it is initially less affected by the DEM errors in flat or near flat areas. The influence matrix approach also yields a picture of the upslope areas contributing to a cell and the downslope areas fed by a cell. This aids in visualizing the flux calculation and helps validate the flux calculation.

To describe the influence matrix algorithm, we need a notation that discriminates between a cell assigned to be the origin of an influence matrix from the same cell when it is used as a point for summing the flux from all upstream influence matrices. When a cell (or related property) is regarded as the start of an influence matrix, we use superscripts. When the cell is regarded as a receiving cell of flux from upstream cells we use subscripts. The contribution of flux from originating cell (l,m) arriving at cell (i,j) is then

$$F_{i,j}^{l,m} = \left(\left(\left(\left(S^{l,m} A \right) \lambda_0 \right) \lambda_1 \right) \dots \right) \lambda_p \right)$$
(16)

Where $S^{l,m}$ is the cell area and $A^{l,m}$ is the accumulation rate of matrix-origination-cell $x^{l,m}$, and $\lambda_0 \dots \lambda_p$ are the partition fractions using the Costal-Cabral and Burges partition scheme (Equation (14) and (15)).

The total flux $(\hat{F}_{i,j})$ through cell $x_{i,j}$ is:

$$\hat{F}_{i,j} = S^{i,j} \stackrel{\bullet}{A}^{i,j} + \sum_{l} \sum_{m} F^{l,m}_{i,j}$$
(17)

where the sums are taken over the cells that have influence matrices that include cell (i,j)

Once fluxes for every cell have been calculated, the balance velocity is computed according to Equation (5).

3.4 Error estimation

Uncertainty in the balance velocity comes from measurement errors associated with the observations and from the behavior of the algorithms. Errors in the behavior of the algorithm can be subtle and they also bias the way we estimate random errors. Here, we begin by discussing the effect of random errors in the observations on the balance

velocity. We go on to discuss how artificial mass diffusion associated with the algorithm also causes errors in the final results.

Random errors in the DEM, accumulation rate and ice thickness are estimated from the data source-documentation. Random errors in the velocity ratio (r) are estimated to be about 10% of our chosen value. These errors propagate when the associated data are used to calculate flow direction, flux and balance velocity. We apply the maximum error propagation theory to estimate the error from input data. Essentially if $y = f(x_1, x_2, ..., x_m)$, and if the maximum errors for $x_1, x_2, ..., x_m$ are $\Delta_{x_1}, \Delta_{x_2}, ..., \Delta_{x_m}$, then the maximum absolute error for $y(\Delta_y)$ is:

$$\Delta_{y} = \left| \frac{\partial f}{\partial x_{1}} \right| \Delta_{x_{1}} + \left| \frac{\partial f}{\partial x_{2}} \right| \Delta_{x_{2}} + \dots + \left| \frac{\partial f}{\partial x_{m}} \right| \Delta_{x_{m}}$$
(18)

The maximum absolute error of flow direction ($\Delta \theta$) is from Equations (9), (10) and (11):

$$\Delta \theta = \left(\frac{\left| E_{i-1,j-1} - E_{i+1,j+1} \right| + \left| E_{i-1,j+1} - E_{i+1,j-1} \right|}{\left(E_{i-1,j-1} - E_{i+1,j+1} \right)^2 + \left(E_{i-1,j+1} - E_{i+1,j-1} \right)^2} \right) \Delta z \tag{19}$$

where Δz is the random DEM error on the difference between neighboring cell elevations.

The absolute error of $\hat{F}_{i,j}$ ($\Delta \hat{F}_{i,j}$) is:

$$\Delta \hat{F}_{i,j} = S^{i,j} \Delta \overset{\bullet}{A}^{i,j} + \sum_{l} \sum_{m} \Delta F^{l,m}_{i,j}$$
(20)

We estimate a maximum error of the last term to be:

$$\Delta F_{i,j}^{l,m} = \left(\lambda_0 \lambda_1 \dots \lambda_p\right) S^{l,m} \Delta \overset{\bullet}{A}^{l,m} + \left(\lambda_1 \lambda_2 \dots \lambda_p\right) S^{l,m} \overset{\bullet}{A}^{l,m} \Delta \lambda_0 + \dots + \left(\lambda_0 \lambda_1 \dots \lambda_{p-1}\right) S^{l,m} \overset{\bullet}{A}^{l,m} \Delta \lambda_p$$
(21)

Similarly, the absolute error ($\Delta\lambda$) of the partition function (λ) (Equations (14) and (15)) is:

$$\Delta \lambda = \frac{1}{2\cos^2 \theta} \Delta \theta \tag{22}$$

Combining Equations (21) and (22) yields:

$$\Delta F_{i,j}^{l,m} = \left(\lambda_0 \lambda_1 \dots \lambda_p\right) S^{l,m} \Delta \overset{\bullet}{A}^{l,m} + \frac{\left(\lambda_1 \lambda_2 \dots \lambda_p\right) S^{l,m} \overset{\bullet}{A}^{l,m}}{\left(2\cos^2 \theta_0\right)} \Delta \theta_0 + \dots + \frac{\left(\lambda_0 \lambda_1 \dots \lambda_{p-1}\right) S^{l,m} \overset{\bullet}{A}^{l,m}}{\left(2\cos^2 \theta_p\right)} \Delta \theta_p \tag{23}$$

.

The relative error of $F_{i,j}^{l,m}$ is:

$$\frac{\Delta F_{i,j}^{l,m}}{F_{i,j}^{l,m}} = \frac{\Delta A}{A} + \frac{\Delta \lambda_0}{\lambda_0} + \dots + \frac{\Delta \lambda_p}{\lambda_p}$$
(24)

 $\Delta F_{i,j}^{l,m}$ includes two characteristic parts. The first is error propagated from observation error in the accumulation rate. The second part includes DEM errors that propagate into computed surface slopes and hence the partitioning factors (λ). The error associated with the partitioning scheme and the flux estimation algorithm propagates through the influence matrix via the terms:

$$\left(\lambda_{1}\lambda_{2}...\lambda_{p}\right)S^{l,m}\overset{\bullet}{A}^{l,m}\Delta\lambda_{0}+...+\left(\lambda_{0}\lambda_{1}...\lambda_{p-1}\right)S^{l,m}\overset{\bullet}{A}^{l,m}\Delta\lambda_{p}$$
(25)

In that sense, Equation (25) implicitly includes diffusion effects, which serve to increase the error estimate at each grid cell.

The error Equations (20-23) also show how errors are accumulated downstream from the ice divide. For a downstream cell, the flux error will increase as *l* and *m* increase (Equation (20)), in other words, error increases as the contributing area increases, or if a smaller grid cell size is chosen for the same area. Equation (23) also shows that when θ (the 0° – 45° angle between flow direction and cardinal direction) is close to 45°, $\Delta F_{i,j}^{l,m}$ is larger than when θ is close to 0°. Restated, the smaller the angle between flow direction and cardinal directioning using Equations (12) and (13) would yield a minimum error at 45° to the grid direction, which we believe is less desirable because the algorithm only allows for flow orthogonal to the grid direction.

We use these results to estimate balance velocity errors. By Equation (5), the absolute value of surface balance velocity (ΔV_s) is:

$$\Delta V_{s} = \frac{1}{H \cdot (\sin\theta + \cos\theta) \cdot W \cdot r} \Delta \hat{F}_{i,j} + \frac{\hat{F}_{i,j}}{(\sin\theta + \cos\theta) \cdot W \cdot r \cdot H^{2}} \Delta H + \frac{\hat{F}_{i,j} \cdot (\cos\theta - \sin\theta)}{H \cdot W \cdot r \cdot (\sin\theta + \cos\theta)^{2}} \Delta \theta$$

$$\dots + \frac{\hat{F}_{i,j}}{(\sin\theta + \cos\theta) \cdot W \cdot H \cdot r^{2}} \Delta r \qquad (26)$$

The fractional error is:

$$\frac{\Delta V_s}{V_s} = \frac{\Delta F_{i,j}}{\hat{F}_{i,j}} + \frac{\Delta H}{H} + \frac{\cos\theta - \sin\theta}{\sin\theta + \cos\theta} \Delta\theta + \frac{\Delta r}{r}$$
(27)

We note again that we assign a single value (0.9) for the ratio between depth-averaged mean velocity and the surface velocity (Equation (5)). Given the range of probable values for r, we assign the error on r to be 0.1.

4. Data sets

We used four primary data sets in this analysis: the OSU DEM of Antarctica (Liu and others, 1999); the BEDMAP (2000) ice thickness model; surface accumulation rates (Vaughan and others, 1999); and RAMP flow stripes.

The elevation data contained in the DEM from Liu and others (1999) are from several topographic data sets that have rather disparate sampling intervals. Consequently, the interpolated product was uniformly resampled to 200, 400 and 1000 m post spacings – recognizing that in some areas this represents an over sampling of the available data. Similarly the quality and accuracy of the data varies due to data collection methods as discussed in Liu and others (1997). For error analysis purposes, we are only concerned with relative error and we take the relative error in elevation (Δz in Equation (19)) between adjacent observations to be 1 m.

Vaughan and others (1999) summarize accumulation rate data. Their product is available as a 10 km cell size, Arc/Info grid. It is assembled from over 1800 published or unpublished in situ measurements. The uncertainty of the data is approximately 10%.

Ice thickness was compiled as part of the BEDMAP project (Lythe and others, 2000). The data is provided in an Arc/Info grid with *5km* cell-size. It is based on about 2 million ice thickness observations by 12 countries over the last five decades. The accuracy of ice thickness is different in different regions and ranges from 10 to 180 meters (Lythe and others, 2000). A few areas, such as the Amery Ice Shelf, Filchner-Ronne Ice Shelf and West Antarctic Ice Streams (WAIS), have good accuracy and excellent coverage (Lythe and others, 2000). Other areas, such as large parts of East Antarctica, are covered by 50km spaced flight-lines or with little to no data at all (there, ice thickness is based on gross interpolation or model results) (Lythe and others, 2000). We estimated ice thickness errors according to known properties of the data collection methods (Figure 3).



Figure 3. Estimated error in ice thickness

Flow stripes in the fast ice streams and ice shelves are evident on the 1997 RAMP Radarsat imagery. As part of a separate project, the stripes were manually digitized and then converted into Arc/Info line coverage. Long, linear features were judged to be flow stripes based on continuity, correlation with known glaciologic features (such as ice streams) and general directions relative to a priori knowledge of surface topography (for example, features running orthogonal to the generally known topography were not identified). Figure 4 shows the digitized flow stripe map from the Radarsat mosaic. Of these, we used in this analysis the flow stripes drawn for: the ice streams draining into the Filchner Ronne and Ross Ice shelves; the Lambert drainage system; Byrd, David and Pine Island Glaciers; and outlet glaciers through the Sor Rondane Mountains. We incorporate the flow line information into the analysis by determining the x-y coordinate of each node making up the line. The line coverage is converted into a line segment coverage and the nodal coordinates of the ends of each segment are used to compute the orientation that is saved as an attribute of each segment. The line segment coverage is then gridded into a 20 km cell where the values of the cell are the orientation angles of the line segments nearest to the center of the cell. The last step is manual inspection and editing (Wu, 2002).



Figure 4. Digitized flow stripes and place names referred to in the text. Abbreviated names are: BIS, Bailey Ice Stream; SFG, Support Force Glacier; WAIS, West Antarctic Ice Streams.

5. Results

Surface balance velocities calculated using the algorithm described in section 3 are shown in Figure 5 on a 20 km grid. Velocities are not calculated over the ice shelves because the flow directions from both the DEM and from flow stripes are suspect. As noted by Bamber and others (2000), the surface balance velocity distribution map shows complexities in flow from the ice divides to the coast. Channeled flow is evident around the entire continent and through major outlet glaciers such as David Glacier in Northern Victoria Land. The map captures the network of tributaries that feed the West Antarctic Ice Streams draining into the Ross and Ronne Ice Shelves. The similarly complex flow of the ice streams draining into the Filchner Ice Shelf is also evident. In particular, the surface balance velocity map depicts the flow of the newly discovered Blackwall Glacier (Jezek, 1999; Gray and others, 2001). Blackwall's twin, the RAMP Glacier, is absent from the balance velocity map even though it appears more clearly in the Radarsat Mosaic (Jezek, 1999). We suspect that the glacier is missed in our model because of the sparse ice thickness data available for this region. Organized flows associated with Recovery Glacier and Support Force Glacier and Foundation Ice Stream snake 100's of kilometers into East Antarctica. Flow associated with Support Force Glacier and Foundation Ice Stream extends nearly to the divide, which partitions flow to the Filchner and Ross Ice Shelves. Just on the opposite side of the divide, tendrils of organized flow descend into Byrd Glacier. Some of the least complex flow appears in Queen Maud Land where a broad lobe of slow moving ice extends towards the coast before the coastal mountains bifurcate the flow into several smaller glaciers.

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Figure 5. Surface balance velocity map (V_s)

Figures 6 shows our estimates of surface balance velocity errors from Equation (26). As might be expected errors on the interior ice sheet are a few tens of m a^{-1} where flow is relatively simple and errors are primarily due to local uncertainties in ice thickness, accumulation rate, surface slope, and the velocity ratio r. As flow is channeled, errors begin to grow and can exceed several hundred m a^{-1} at the mouths of outlet glaciers and ice streams.



Figure 6. Surface balance velocity error estimate (ΔV_s)

Figure 7 is the percentage error in surface balance velocity. The blockiness in Figure 7 is caused by the regional assignments of errors in ice thickness. Percentage errors are less than about 40% over much of the ice sheet where the primary contributors to error are local accumulation rate, ice thickness and velocity ratio (r) uncertainties. Errors are largest in regions of converging flow and where the flow approaches an angle of about 45° to the cardinal direction.



Figure 7. Percentage errors in balance velocity $(\Delta V_s / V_s)$

6. Validation and discussion

We compare the surface balance velocity result with independently measured velocity to validate our calculation. We also compare our influence matrix approach to our implementation of the DEM sorting algorithm. Our sorting algorithm calculates flux from the highest to lowest elevation cell, partitioning the flux between cells using Equations (14) and (15). The direction angle for flux partitioning from each cell is based on the flow-stripe direction (where available) rather than the direction from the DEM slope. This is a modification to the DEM sorting approach described by Budd and Warner (1996). By including the flow stripe information in the algorithm, we introduce

an inconsistency in that the DEM is assumed accurate for elevation sorting. By changing the flow direction from slope to flow stripes, we are implying that there are DEM errors that in turn imply that the sort order may be wrong. For the influence matrix approach, we start with the DEM and calculate flow directions. We then essentially discard the DEM elevation data. We finally make modifications to the flow field using the flow stripes. We think this is a physically more consistent approach.

6.1 Lambert Glacier

Global Positioning System (GPS) derived velocities were collected along a profile around the Lambert Glacier drainage system as part of the Australian National Antarctic Research Expedition (Figure 8). The velocities are available through the National Snow and Ice Data Center web site [http://nsidc.org/data/velmap/amery/amery.html] and we compare these to our balance velocity results.



Figure 8. Lambert Glacier and Amery Ice Shelf. Repeat GPS velocity measurement sites are shown by the black dots. Flow lines are based on interpretation of the Radarsat mosaic.

Surface balance velocities are resampled for comparison with the GPS derived velocity data (V_{GPS}) using the nearest grid cell interpolation method from the original 20-km cell-size surface balance velocity result. The results are compared in Figure 9 and Table 1. The mean difference between the influence matrix model $(V_{s_{-}in})$ and the measured result is 2.1 m a⁻¹ with standard deviation of 9.2 m a⁻¹. The mean difference between balance velocity using DEM sorting algorithm $(V_{s_{-}ds})$ and V_{GPS} is 2.5 m a⁻¹ with standard deviation of 9.7 m a⁻¹. The mean difference between balance velocity using the influence matrix algorithm without using flow stripes to correct flow directions $(V_{s_{-}no})$ and V_{GPS} is 2.7 m a⁻¹ with standard deviation of 10.5 m a⁻¹. In this case, all the approaches perform about equally presumably because the DEM (and hence the derived flow directions) as well as the other observational data are accurate in this well studied region. If the DEM

was inaccurate, we might expect an improved result when flow stripe information was added to the mix.

As illustrated in Figure 9, the V_{s} in estimates generally agree with the V_{GPS} data to within the calculation errors. Model and measurements deviate somewhat at locations A, B and C-D. The lateral scales of the anomalies at A and B are only one or two grid cells and we suspect our model simply does not have the resolution to capture these features. Anomalies between C and D are large in lateral scale (relative to the grid size) and in the overall negative difference between measured and modeled velocities (relative to the estimated errors). Points C and D bound the flow of a central tributary feeding Lambert Glacier. The negative differences between measured and modeled velocities suggest that this area is thickening. Previous to our analysis, Fricker and others (2000) compared measured and modeled balance fluxes for the Lambert Drainage Basin. They used a DEM sorting approach to model the flux computed on a 5 km grid. They conclude as we that the flow outside the band between about 800 and 1500 km on Figure 9 is in balance to within the accuracy of the calculation. Our result suggests that the 'stream' region between 800 and 1500 km is also in balance save for the narrow zone bounding the central tributary of Lambert Glacier (C-D). We note however that our balance velocity errors may be significantly underestimated if, as Fricker and others suggest, the accumulation rate uncertainties greatly exceed our assigned value of 10%.

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	Mean Difference (m a ⁻¹)	Standard Deviation (m a ⁻¹)
V_{s_in}	2.1	9.2
V_{s_ds}	2.5	9.7
V _{s_no}	2.7	10.5

Table 1. Surface balance velocity comparisons with GPS velocity (V_{GPS}) for the Lambert Glacier Drainage area.



Figure 9. Measured (solid line) and calculated (black squares) velocity profiles around Lambert Glacier. Errors in measured velocities are less than 1 m a⁻¹.

6.2 Ice Streams Draining into the Filchner Ice Shelf

Zhao (2001) used Radarsat interferometric data to measure surface velocities of ice streams draining from East Antarctic into the Filchner Ice Shelf (Figure 10)



Figure 10. Ice Streams draining into the Filchner Ice Shelf and profile line used to compare balance velocities to InSAR velocities. Abbreviated names are: FIS, Filchner Ice Shelf; BIS, Bailey Ice Stream; RG, Recovery Glacier; BW, Blackwall Ice Stream.

The resulting velocity fields were compiled into a 200-m cell grid with an estimated speed accuracy of 15 ma^{-1} . We compared the InSAR velocities to surface balance velocities calculated using the nearest grid cell interpolation method applied to the original 20-km cell-size surface balance velocity result. The dashed line in Figure 10 shows the profile location and Figure 11 shows the results. Note the improved agreement between surface balance velocities and measured velocities when flow stripe information is used in the analysis. In Table 2, the mean difference between surface balance velocity using the influence matrix algorithm ($V_{s_{in}}$) and interferometric velocity (V_{inSar}) is 3.5 m a⁻¹ with standard deviation of 48.9m a⁻¹. The mean difference between balance velocity

using the DEM sorting algorithm (V_{s_ds}) and V_{lnSar} is 10.0 m a⁻¹ with standard deviation of 45.9 m a⁻¹. The mean difference between InSAR velocities and balance velocities calculated using the influence matrix algorithm but without using flow stripes to correct flow directions (V_{s_no}) and V_{lnSar} is 13.8 m a⁻¹ with standard deviation of 71.9 m a⁻¹.

	Mean Difference (m a ⁻¹)	Standard Deviation (m a ⁻¹)
V _{s_in}	3.5	48.9
V_{s_ds}	10.0	45.9
V _{s_no}	13.8	71.9

Table 2. Surface balance velocity comparisons with interferometry velocity (V_{InSar})



Figure 11. Measured (thick line) velocity and surface balance velocity profiles across ice streams draining into the Filchner Ice Shelf. Influence matrix result with flow stripe information is shown by the black squares. The short dashed line shows the influence matrix result without flow stripe information. DEM sorting approach with flow stripe information is shown with the long dashed line.

There is generally good agreement between the InSAR observations and the calculated balance velocities. Point A corresponds to the inferred margin of Bailey Ice Stream, which Zhao reported to be thickening by 0.25 ± 0.06 m a⁻¹. The increasing difference between balance velocities and InSAR velocities from the start of the profile to about 30 km past point A weakly supports Zhao's observation. The peak flux from Slessor Glacier is located at point C and there is little significant difference between measured and modeled velocities, suggesting, as did Zhao, that Slessor Glacier is in equilibrium. The profile intercepts the northern margin of Recovery Glacier at point D. The balance

velocities are statistically smaller than the measured velocities near the center of the ice stream, again weakly supporting Zhao's calculation which indicates that Recovery Glacier is thinning by 0.23 ± 0.22 m a⁻¹. However we are cautious about drawing a firm conclusion about Recovery Glacier because of the very limited ice thickness data available for this region.

As a further comparison, we have overlaid the measured flow stripes onto the balance velocity map in Figure 12. As noted, most of the flow stripes are used in the balance velocity calculation and in that sense the comparison serves as a consistency check. Shorter flow stripes associated with glaciers in Queen Maud Land and Wilkes Land were not used in the calculation. Although the orientations of these stripes match well with the patterns of balance velocity, the balance velocity data indicate organized flow further into the interior than would be suggested by the flow stripes, for example the flow from Totten Glacier and around Law Dome, the flow down Byrd Glacier and the flow from Support Force Glacier and Foundation Ice Stream.



Figure 12. Flow stripes and balance velocity map (gray image with relatively faster speeds shown lighter than the darker relatively slower speeds)

7. Conclusion

We have developed a modified surface balance velocity estimation approach. Our matrix approach allows us to visualize upslope source and downslope outflow areas associated with individual cells, which helps to validate results. It also facilitates error estimation and can be used to illustrate how errors from source data and biases from the form of the algorithm introduce uncertainty into the result. Our approach flexibly incorporates refined flow directions derived from image data in low slope or complex slope. We find favorable comparison with independent velocity measurements around Lambert Glacier and Ice Streams draining into the Filchner Ice Shelf. The comparisons suggest that both glacial regimes are close to equilibrium within the estimated errors. The computer code version of our balance velocity model is available from X. Wu.

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