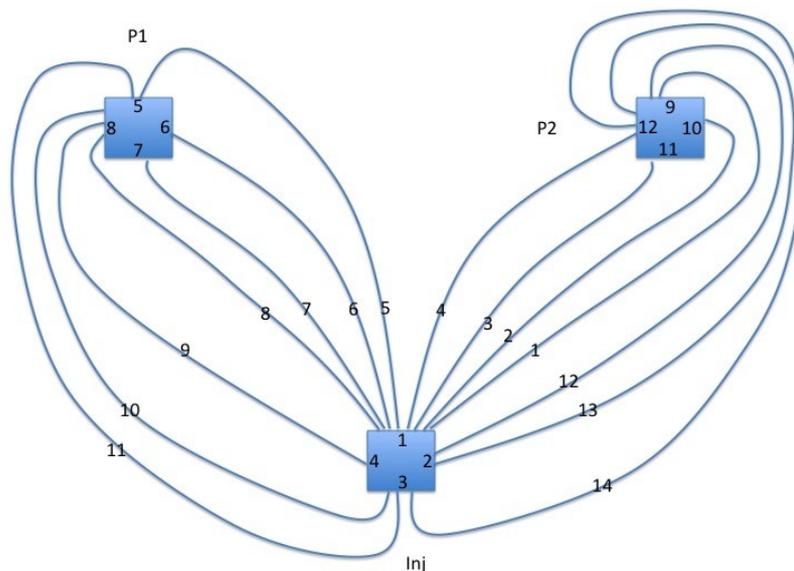


Optimizing streamline flux assignments (fluxopt)

The example

The configuration we looked at is as in the figure. It has 3 wells (one injector and two producers) in 2D. Shown are 14 streamlines. Note that these are really streamline clusters: each face-to-face connection is represented by only one streamline. In reality there will be more streamlines per cluster. In this setting, with the chosen face-to-face connections, the maximum number of unique clusters is the given 14.



General overview of the approach

Let's call the total number of well faces n_{wf} , the faces are denoted by $f(i)$, the streamlines by j , $j=1, \dots, n_{sl}$, and the streamline fluxes by $slf(j)$.

We assume that each $f(i)$ has a unique total flux $F(i)$ in or out of the face (so, we are not assuming that for each well, the total well rate is equally divided over its faces).

We also assume that each well face has streamlines exiting or entering it (which will be the case in practice as all grid cells are required to be crossed by at least one streamline).

To simplify the calculations we assume that all fluxes, be it well face fluxes or streamline fluxes, are positive. When needed, we negate producer fluxes in the system of equations.

For each well face $f(i)$, denote by S_i the set of streamlines j that start or end at this well face. For each face $f(i)$, we set up an equation of the form

$$\sum_{j \in S_i} slf(j) = F(i)$$

This leads to a matrix-vector system $Ax=b$, with the vector x containing the $slf(j)$'s

and the vector b the fluxes $F(i)$.

The rank of the matrix A will always be equal to the $nwf-1$. It is not nwf because any one of the face fluxes is in fact dependent on the others (it is the total reservoir flux – sum of the other $nwf-1$ well face fluxes). The right hand side b has that dependency in it as well of course.

What this means is that $Ax=b$ will always have an exact solution. However, this solution is not useful in general as it likely returns several negative $slf(j)$.

In order to solve $Ax=b$ optimally, we need to add positivity constraint for the solutions $slf(j)$. This can be done, for example, by `lsqnonneg` in MATLAB. Using a non-negative least squares method such as this also easily allows additional constraints to be added, or constraints to be weighted nonuniformly.

What's the procedure?

1. Set up $Ax=b$
2. Perform a QR to find the relevant streamline clusters (this bundles all streamlines that go from the same face to the same face. In 2D, such streamlines will be adjacent to each other on a face. In 3D they will likely be in the same local area, but may not be fully adjacent)
3. Solve the reduced system for the streamline cluster fluxes.
4. Knowing the total flux for each streamline cluster, now divide that flux over the streamlines in each cluster.

Note that in step 4, we have complete freedom. It could be done to ensure that neighboring clusters (eg one on one face close to a corner and the other just around the corner on the next face) transition smoothly. This will, we think, reduce mapping errors a great deal (because then we avoid having sudden changes in volume in streamlines that leave just around the corner from each other and can end up in similar cells).

Adding constraints

The above method tries to match the face fluxes optimally. The end result does not guarantee, of course, that the total volume fluxes for individual injectors or producers or satisfied only that the well face fluxes are matched optimally.

However, the method allows the addition of constraints. For example, we can add a total flux constraint for each well. We can also add an equation that gives the total flux of all injectors together (which would force total volume balance). Adding these additional constraints will reduce the mismatches for the total well fluxes.

Weighting your favorite constraints

So far, we have not given any more weight to one constraint or another. The methods allows us to do this however.

We can create a diagonal matrix W and instead find the best nonnegative solution to $WAx= Wb$. Here w_{ii} , the i th diagonal element of W , gives the weight imposed on constraint i . For example, if we want to create a solution that preserves total volume

injected and constraint k is the corresponding added constraint, a large w_{kk} relative to the other w_{ii} 's will force this constraint to be satisfied.

Preliminary tests

We now consider two test cases for the example problem.

Case 1:

Here, each face of a well carries a quarter of the total well flux. The total injector flux is 100, the total flux of producer 1 is 40 and producer 2 has a flux of 60.

We compare

1. The current 3DSL procedure of distributing the injector face fluxes equally amongst the streamlines leaving each face
2. The non-negative least squares procedure (lsqnonneg) with equations for each well face and no additional constraints
3. lsqnonneg with the additional constraint for total injector flux
4. lsqnonneg with the additional constraints for total injector and producer fluxes
5. Same as 4 with the additional constraints weighted 100 times stronger than the others.

Fluxes for each streamline are (oops, I don't know where the first numbers in the last column went...)

streamline	(1)	(2)	(3)	(4)	(5)
1	3.1250	0	0	0	
2	3.1250	10	10.3714	11.0281	
3	3.1250	10	10.3714	11.0281	
4	3.1250	0	0	0	
5	3.1250	0	0	0	
6	3.1250	5	5.3714	4.8007	
7	3.1250	5	5.3714	4.8007	
8	3.1250	0	0	0	
9	25	17.5	18.4286	18.2292	
10	8.333	0	0	0	0
11	8.333	12	12.7429	12.0924	12.1053
12	12.5	17	18.3197	18.3197	18.6842
13	12.5	6	6.0462	6.0462	6.0526
14	8.333	11	12.2736	12.2736	12.6315

The corresponding residuals (assigned – actual)

Well-face	(1)	(2)	(3)	(4)	(5)
1	0	5.0000	6.4857	6.6576	7.1052
2	0	-2.0000	-0.8857	-0.6341	-0.2632

3	0	-7.5000	-6.5714	-6.7708	-6.5790
4	1.4583	2.0000	2.7429	2.0924	2.1053
5	-6.8750	-5.0000	-4.6286	-5.1993	-5.2631
6	-6.8750	-5.0000	-4.6286	-5.1993	-5.2631
7	26.4583	7.5000	8.4286	8.2292	8.4210
8	0.6250	2.0000	2.7429	3.3197	3.6842
9	-11.8750	-5.0000	-4.6286	-3.9719	-3.6842
10	-11.8750	-5.0000	-4.6286	-3.9719	-3.6842
12	8.9583	2.0000	2.7429	3.3197	3.6842
Inj	0	-6.5	-1.8571	-1.3813	-0.0002
P1	14.1667	-0.5	1.9143	-0.0770	<0.0001
P2	-14.1667	-6.0	-3.7714	-1.3043	-0.0002

Clearly, we can make the method volume conservative (see results for 5). There are of course mismatches for the faces, but not in the total volume coming in to each producer.

It is possible with weights to, for example, make sure that important faces are well represented.

Case 2:

The injector has face data (70,10,10,10) and the two producers are as in case 1. This case is a little more realistic because F(1) is large, and f(1) indeed has most streamlines emanating from it.

In this case the exact solution is in fact non-negative and all the variants of the method yield the same result, listed in the second column below. The first column is again the 3DSL procedure.

Streamline fluxes are

streamline	(1)	(2)
1	8.7500	10.0000
2	8.7500	15.0000
3	8.7500	15.0000
4	8.7500	0
5	8.7500	10.0000
6	8.7500	10.0000
7	8.7500	10.0000
8	8.7500	0
9	10.0000	10.0000
10	3.3333	0
11	3.3333	0
12	5.0000	5.0000
13	5.0000	5.0000
14	3.3333	10.0000

with residuals for each well-face and the residual in the total flux of the injector, (Inj)

and the residual in the total flux of the producer (P1,P2) in the following table

Well-face	(1)	(2)
1	0	0.0000
2	0	-0.0000
3	0	-0.0000
4	0	0.0000
5	2.0833	-0.0000
6	-1.2500	0.0000
7	-1.2500	0
8	12.0833	0.0000
9	-1.2500	0.0000
10	-6.2500	0.0000
11	-6.2500	-0.0000
12	2.0833	0.0000
Inj	-0.0000	0.0000
P1	11.6667	0.0000
P2	-11.6667	0.0000

We get a REALLY NICE SOLUTION!!!!!!!!!!

Since this seems to be a more realistic example in which a larger number of streamlines corresponds with a larger face flux we are pretty optimistic that in a real case this is going to work really well.

Next steps are to

1. look at a larger realistic case
2. return clusters as well as cluster flux assignments
3. distribute the cluster fluxes over the cluster streamlines
4. compare actual solutions and mb errors