



INTRODUCTION TO GEOPROCESSING CONFLATION TOOLS AND WORKFLOWS

Dan Lee

dlee@esri.com

Agenda

What is Conflation?

Geoprocessing Conflation Tools

- **Demo 1 – Basic scenario**

Conflation Workflows

- **Demo 2 – Real world scenario**

Conclusions and Future Work

What is Conflation?

Translated by Esri localization

Zusammenführung

Fusione F^1/E

合并 *Объединение*

補正 *Combinação*

Combinación

Birleştirme

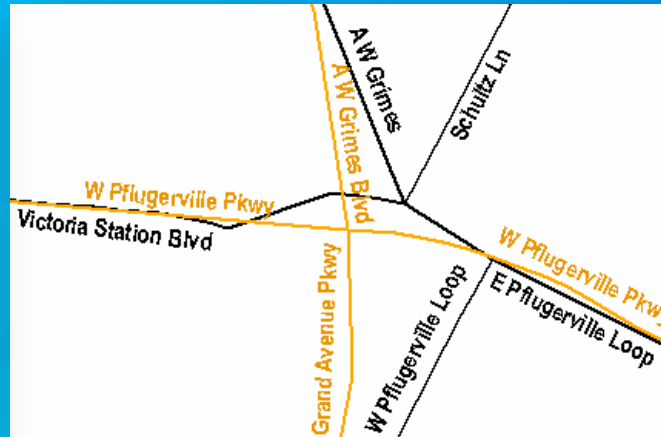
Assemblage

When using multi-source spatial data **together**

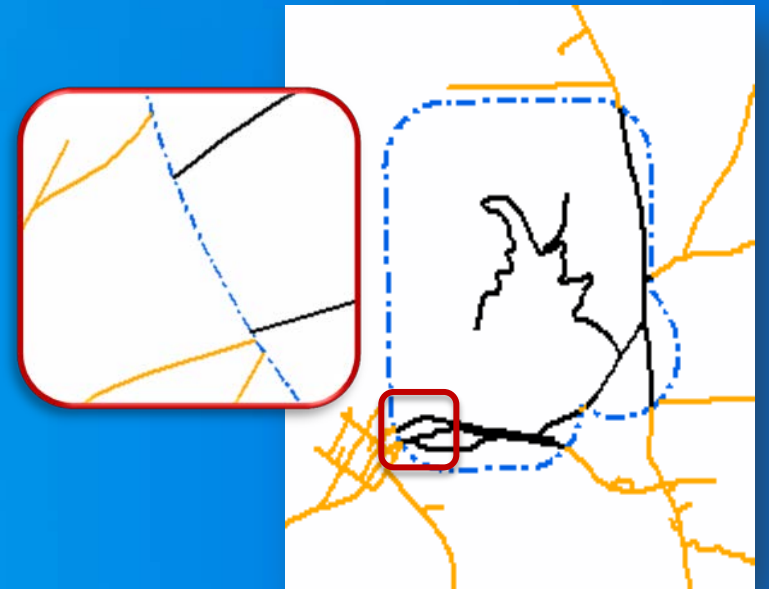
Common obstacles in analysis and mapping:

- **Spatial and attribute inconsistency caused by differences in data collection and modeling**
- **High cost to fix the problems**

Overlapping datasets



Adjacent datasets



Conflation reconciles multi-source datasets and optimizes data quality and usability

Conflation is the process of:

- Identifying corresponding features (known as feature matching)
- Making spatial adjustment and attribute transfer
- Ultimately, combining matched and unmatched features into one unified dataset with the optimal accuracy, completeness, consistency, and integrity

Long-term benefits:

- No longer living with various imperfect datasets
- More confidence in reliable analysis and high quality mapping

What's the way to get there?

The background features a blue gradient with abstract geometric shapes in purple and yellow on the left side. A small yellow map fragment is visible within one of the shapes.

Geoprocessing Conflation Tools

Our initial focuses

Develop highly automated tools in Geoprocessing framework:

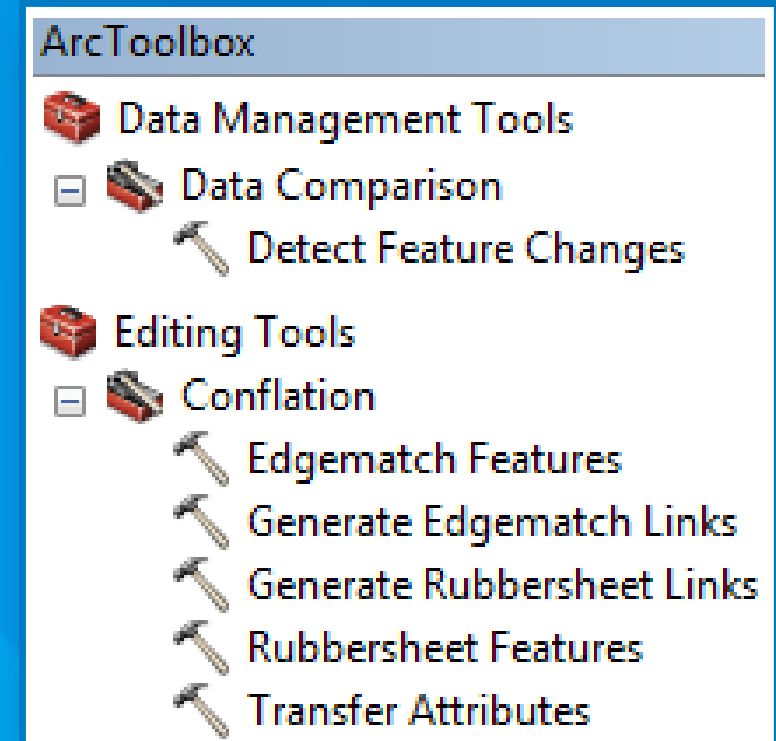
- Starting with linear features (roads, parcel lines, etc.)
- Aiming at high feature matching accuracy (not promising 100%)
- Providing information to facilitate post-processing

Build practical workflows

Have you used these tools?



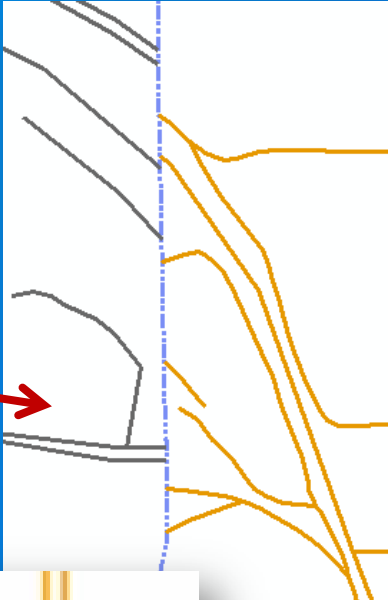
New in ArcGIS 10.2.1



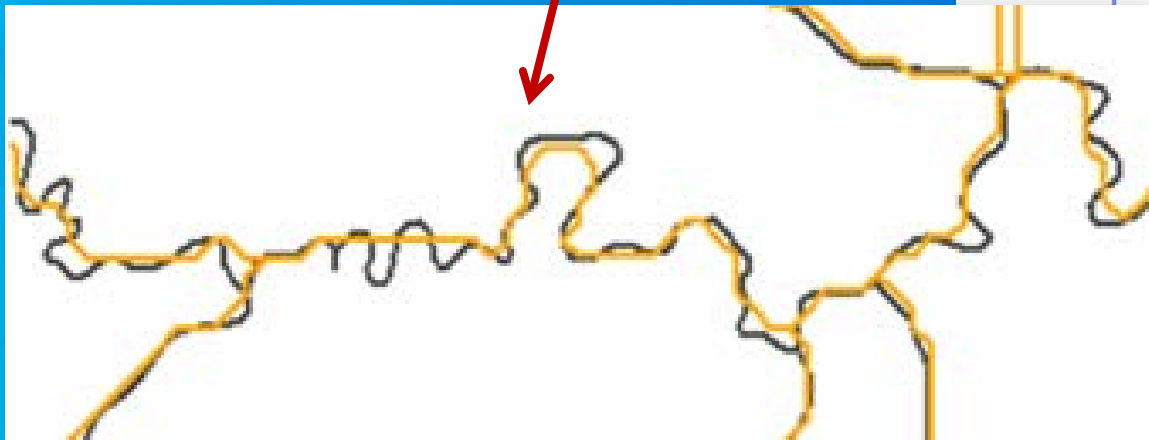
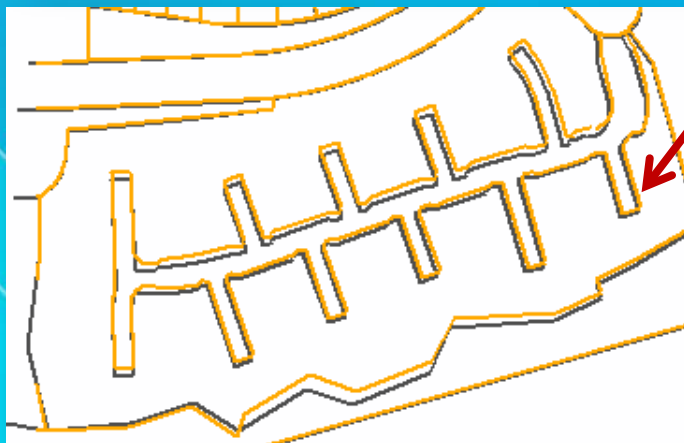
Challenges in feature matching - the key to conflation



Complexity

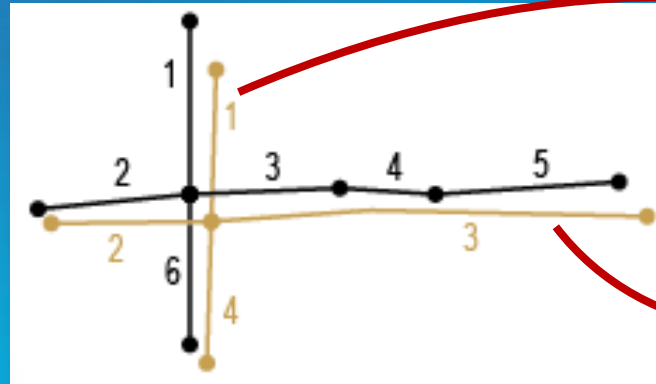


Dissimilarity



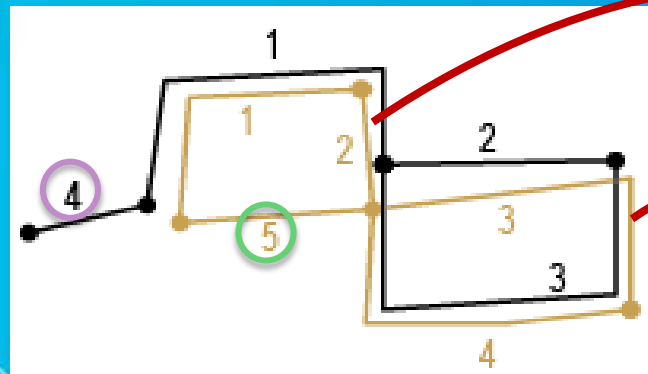
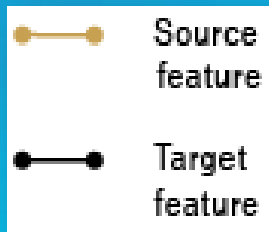
Feature matching (FM) for overlapping datasets

Based on proximity, topology, pattern, and similarity analysis, as well as attributes information



1:1 and 1:m matches

OBJECTID +	SRC_FID	TGT_FID	FM_GRP	FM_MN	FM_CONF
1	1	1	1	1:1	100
2	2	2	2	1:1	100
3	3	3	3	1:3	100
4	3	4	3	1:3	100
5	3	5	3	1:3	100
6	4	6	4	1:1	100



m:1 and m:n matches

OBJECTID +	SRC_FID	TGT_FID	FM_GRP	FM_MN	FM_CONF
1	1	1	1	2:1	100
2	2	1	1	2:1	100
3	3	2	2	2:2	100
4	3	3	2	2:2	100
5	4	2	2	2:2	100
6	4	3	2	2:2	100
7	5	-1	-1	N/A	0
8	-1	4	-1	N/A	0

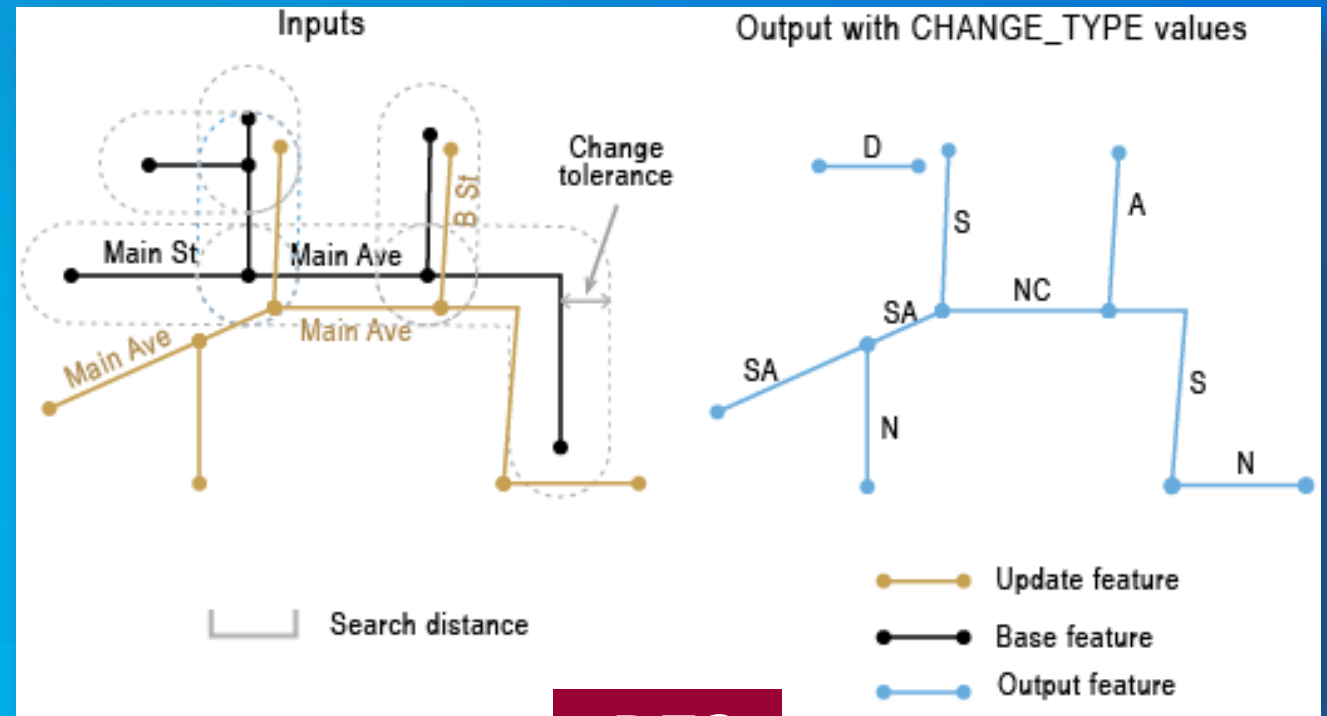
FM-based tool #1 - Detect Feature Changes (DFC)

Finding feature differences

Update features vs.
base features

Output CHANGE_TYPE

- Spatial (S) change
- Attribute (A) change
- Spatial and attribute (SA) change
- No change (NC)
- New update feature (N)
- To-Delete base feature (D)

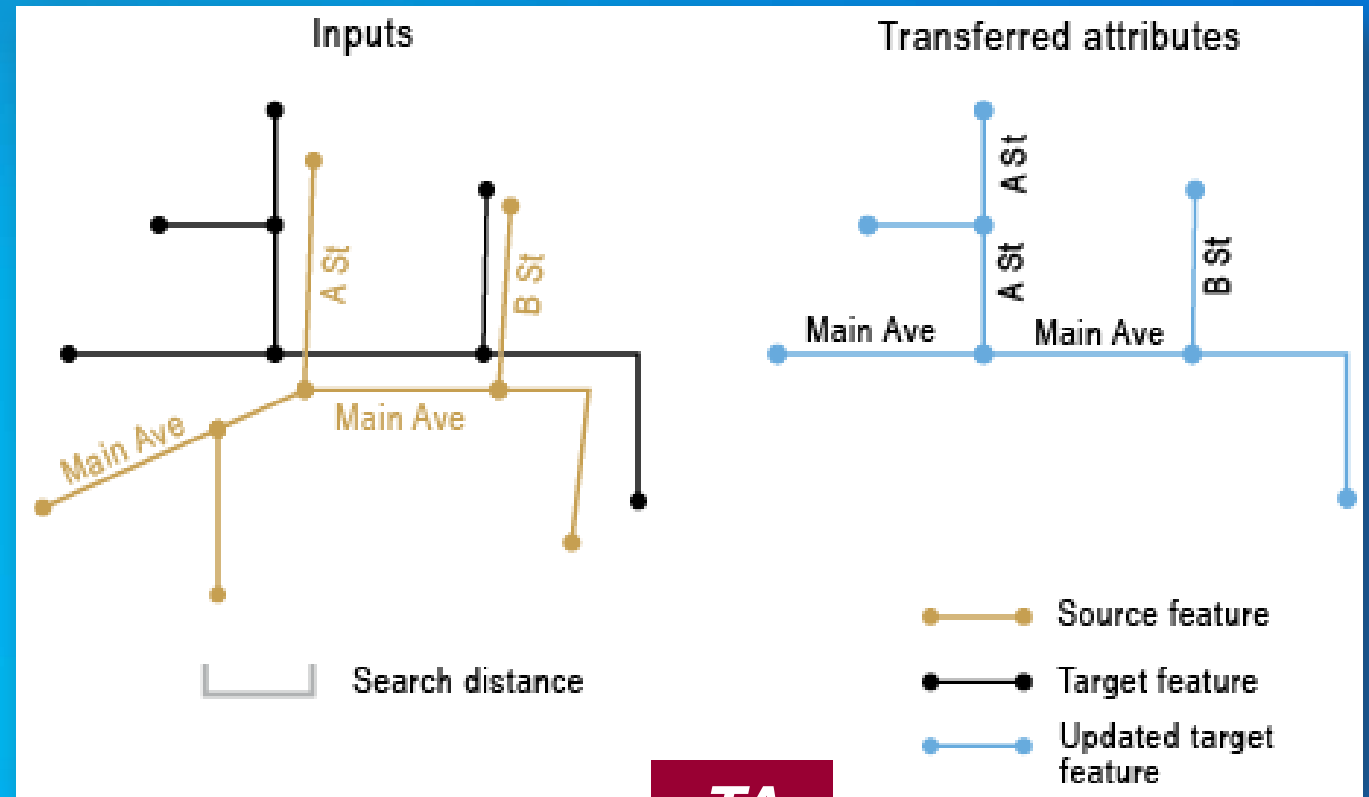


DFC

FM-based tool #2 – Transfer Attributes (TA)

From source features to target features

- Transfer fields (e.g. ROAD_NAME)
- Target features are modified



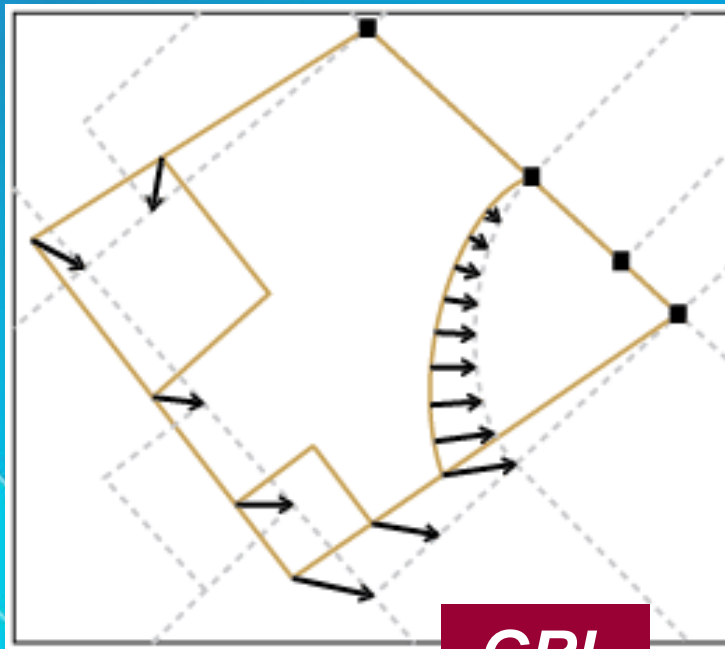
TA

FM-based tool #3 – Generate Rubbersheet Links (GRL)

Rubbersheeting moves source locations towards target locations based on established links

Generate Rubbersheet Links (GRL)

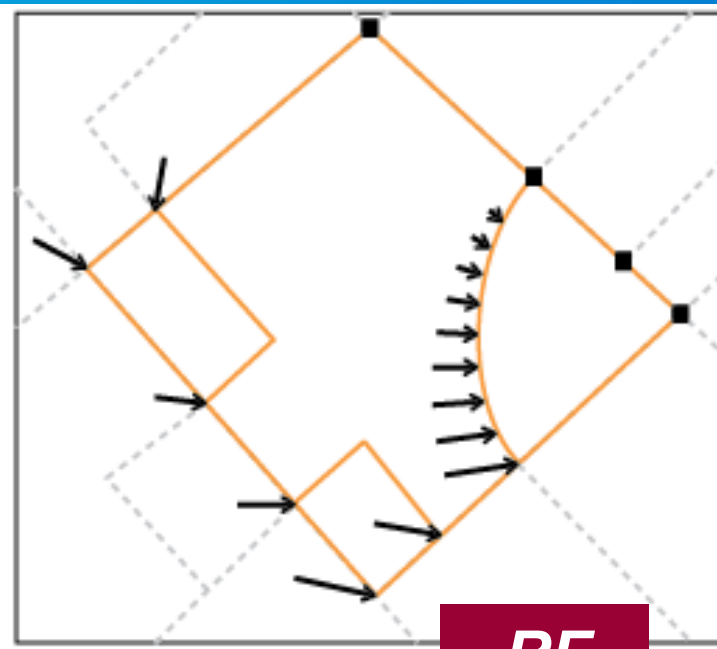
- From source features to target features



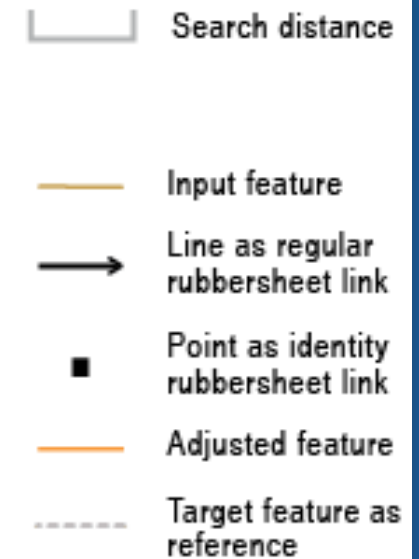
GRL

Followed by Rubbersheet Features (RF)

- Adjusting input features



RF



Edgematching (EM) for adjacent datasets

Based on proximity, topology, and continuity analysis, as well as attributes information

Generate Edgematch Links (GEL)

- From source features to adjacent features

Followed by Edgematch Features (EF)

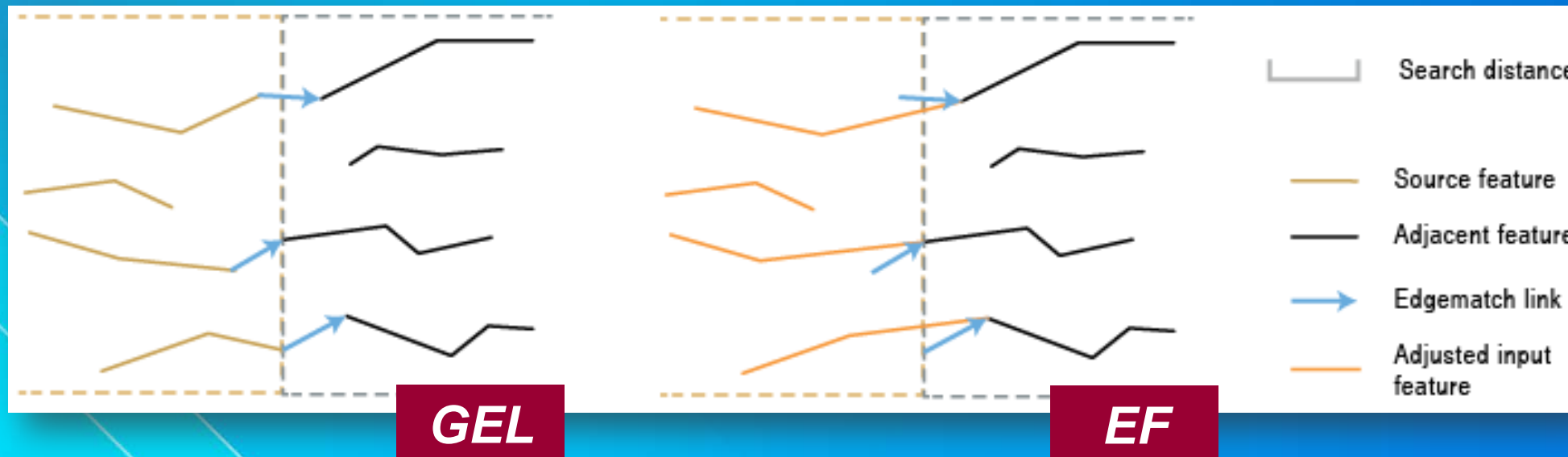
- Connects features guided by the established links

Recommend:

Conflation: Edgematching tools and workflows

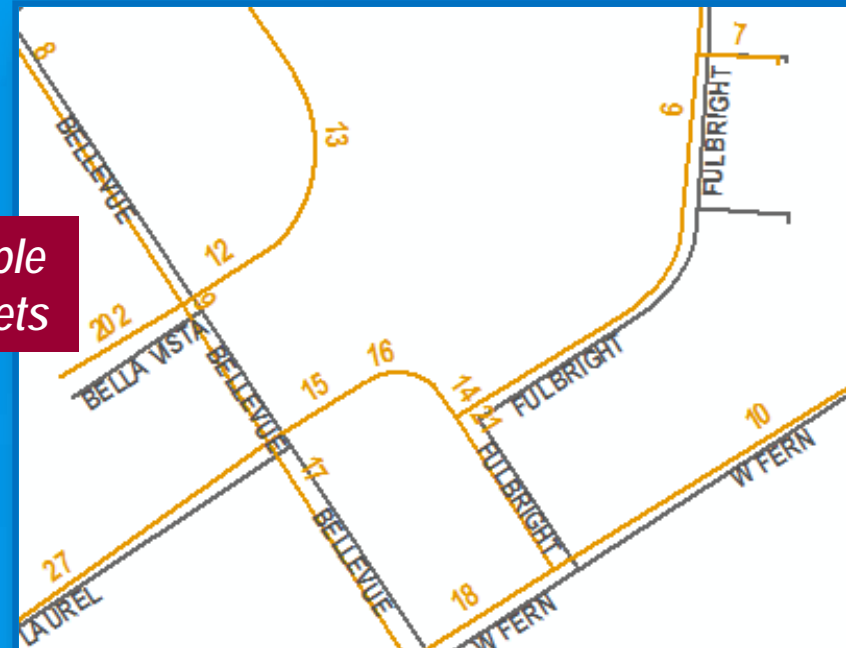
12:30 – 1:15pm, Thur.

Demo Theater – Analysis & Geoprocessing, Hall B



Demo 1: Basic scenario

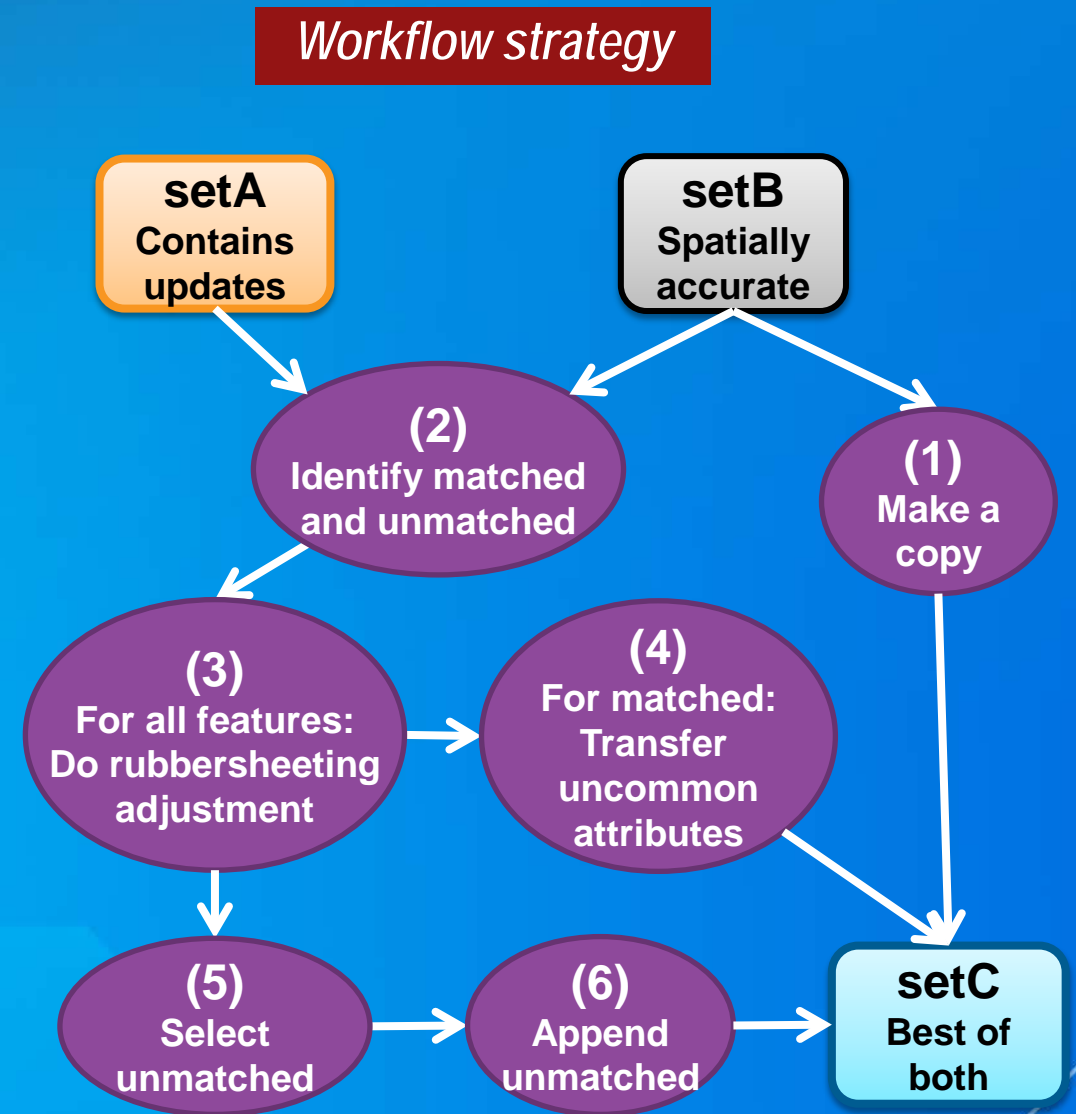
*Unification of simple
overlapping datasets*



Unification of overlapping datasets

A popular scenario and requirements:

- To unify the two datasets into one with combined spatial and attribute information



Input streets

Together

Update features with new streets and attributes

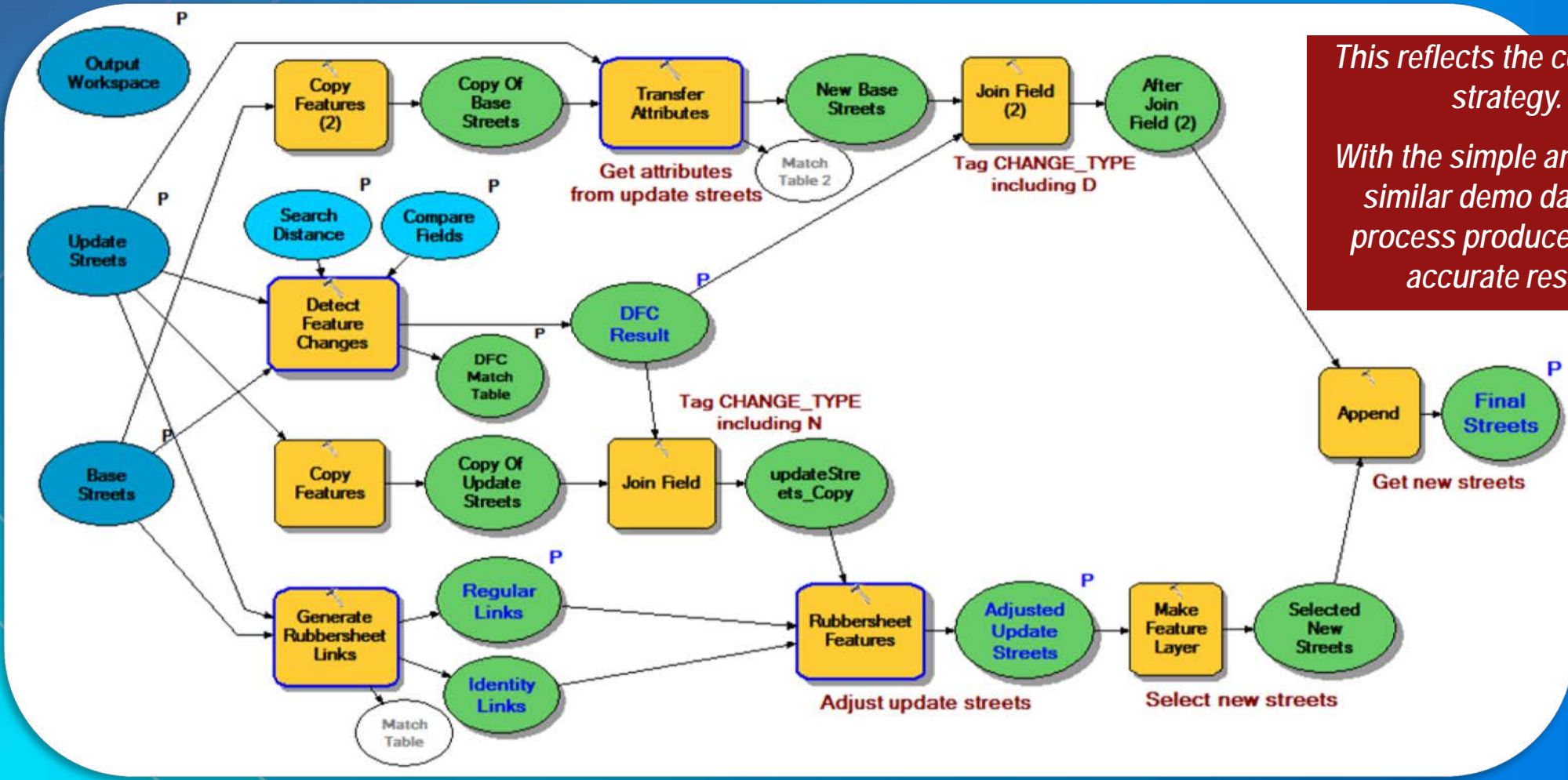
OBJECTID *	Shape	UniqueID	Priority	Shape_Length
1	Polyline	1	<Null>	88.544127
2	Polyline	2	L	298.351168
3	Polyline	3	L	166.198499
4	Polyline	4	L	83.759794
5	Polyline	5	L	109.060742
6	Polyline	6	L	378.689485
7	Polyline	7	L	44.406315
8	Polyline	8	M	178.995089
9	Polyline	9	M	82.505747
10	Polyline	10	M	326.699698
11	Polyline	11	<Null>	113.142624
12	Polyline	12	<Null>	59.711738
13	Polyline	13	<Null>	90.582727
14	Polyline	14	<Null>	18.331248



Base features with spatial accuracy and attributes

OBJECTID *	Shape *	NAME	Priority	Shape_Length
130	Polyline	W FERN	M	82.64188
149	Polyline	W FERN	M	316.419224
156	Polyline	W FERN	M	87.413082
106	Polyline	SHERWOOD	L	138.32864
129	Polyline	SHERWOOD	L	154.974493
73	Polyline	SAN RAFAEL	L	94.960556
82	Polyline	SAN RAFAEL	L	86.952636
87	Polyline	SAN PABLO	L	259.025112
102	Polyline	SAN PABLO	L	162.609722
114	Polyline	NOTTINGHAM	L	138.470439
140	Polyline	NOTTINGHAM	L	217.51654
66	Polyline	MILLS	L	90.060797
69	Polyline	MILLS	L	91.095418

Part 1 - Conflate overlapping datasets



This reflects the conflation strategy.

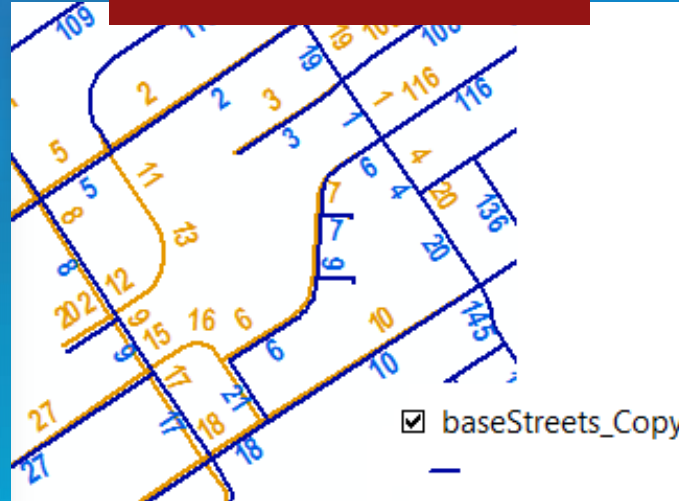
With the simple and highly similar demo data, the process produces 100% accurate result.

Results

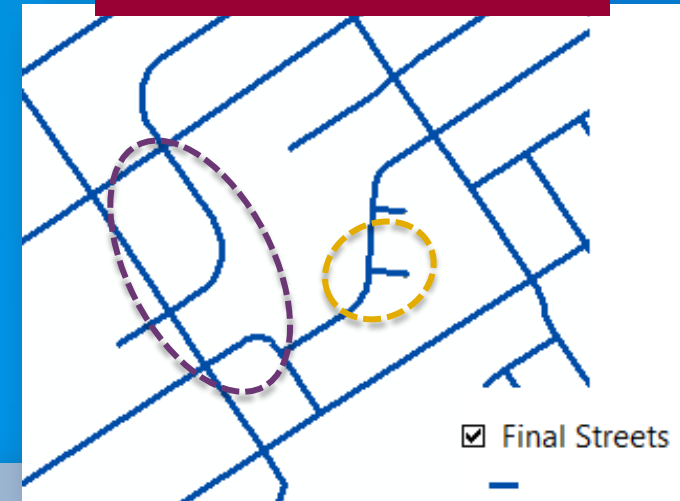
Changes detected



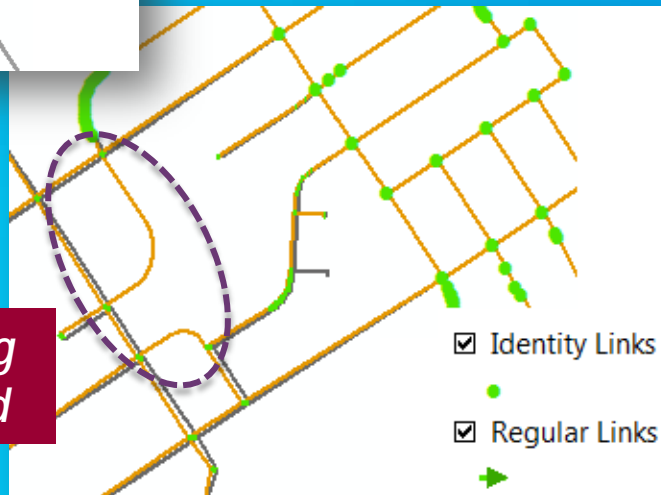
Attributes transferred



New features adjusted and added to base



Rubbersheeting links generated



Final Streets

OBJECTID *	Shape *	NAME	Priority	Shape_Length	UniqueID	CHANGE_TYPE
100	Polyline	W OLIVE	M	293.577825	2	A
109	Polyline	W OLIVE	M	109.071651	5	A
128	Polyline	<Null>	L	51.455286	<Null>	D
200	Polyline	<Null>	M	112.706047	11	N
201	Polyline	<Null>	M	59.242967	12	N
202	Polyline	<Null>	M	90.29932	13	N
203	Polyline	<Null>	M	18.383534	14	N
204	Polyline	<Null>	M	58.252814	15	N
205	Polyline	<Null>	M	37.23	16	N
6	Polyline	LAUREL	L	384.916164	27	NC
19	Polyline	W FERN	M	340.788763	40	NC
66	Polyline	MILLS	L	90.060797	87	NC
67	Polyline	KATHL	L	173.068809	88	NC
69	Polyline	MILLS	L	91.095418	90	NC



Conflation Workflows

Three components in conflation workflows



- In same projection
- Data validation
- Selection of relevant features



- Conflation tools
- Workflow tools



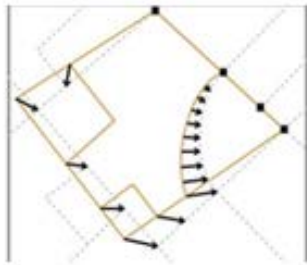
- Queued review
- Interactive editing

Supplemental tools and guidelines for download

<http://angp.maps.arcgis.com/home/item.html?id=36961cde1b074f1f944758f6abec87cc>

You can also search by “conflation” at arcgis.com to find the download.

GP Conflation_Workflow_Guidelines



This item contains conflation workflow guidelines and supplemental tools.

 Geoprocessing Sample by 1812

Last Modified: July 14, 2015

☆☆☆☆☆ (0 ratings, 8 downloads)

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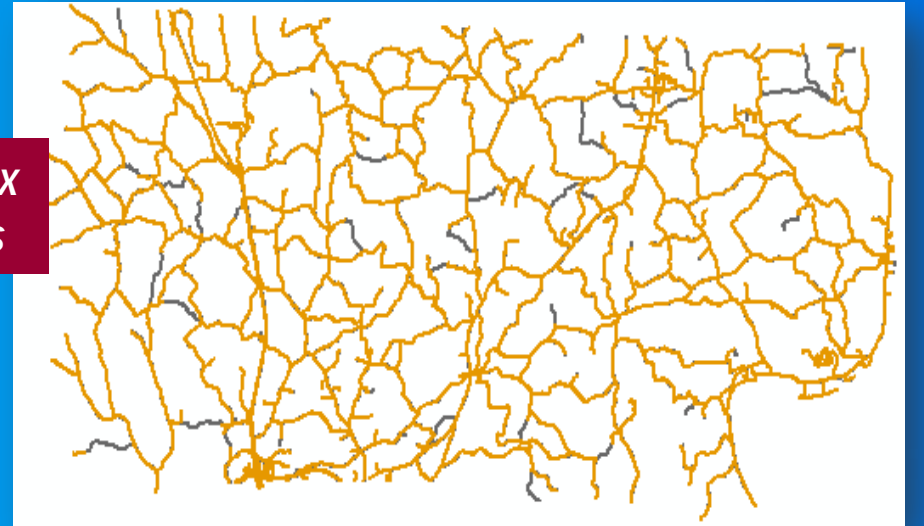
 Facebook  Twitter

OPEN ▾

- Conflation Workflow Tools.tbx
 - Support Tools
 - Check DFC D N NC Types
 - Check Feature Matching
 - Check TA No Value
 - Extract And Classify Feature Vertices
 - Flag Vertex Type Difference on Rubbersheet Links
 - Make Histogram
 - Report DFC Results
 - Workflow Steps
 - Step1a DFC and Evaluation
 - Step1b Extract Matched Features
 - Step2 GRL and Evaluation
 - Step2 GRL and Evaluation for Erik
 - Step3 Rubbersheeting and Assessment
 - Step4 TA and Evaluation
 - Step5 Append N For Final
 - Step6a GEL and Evaluation
 - Step6b Update Link Info
 - Step7 Edgematch

Demo 2: Real world scenario

*Unification of complex
overlapping datasets*



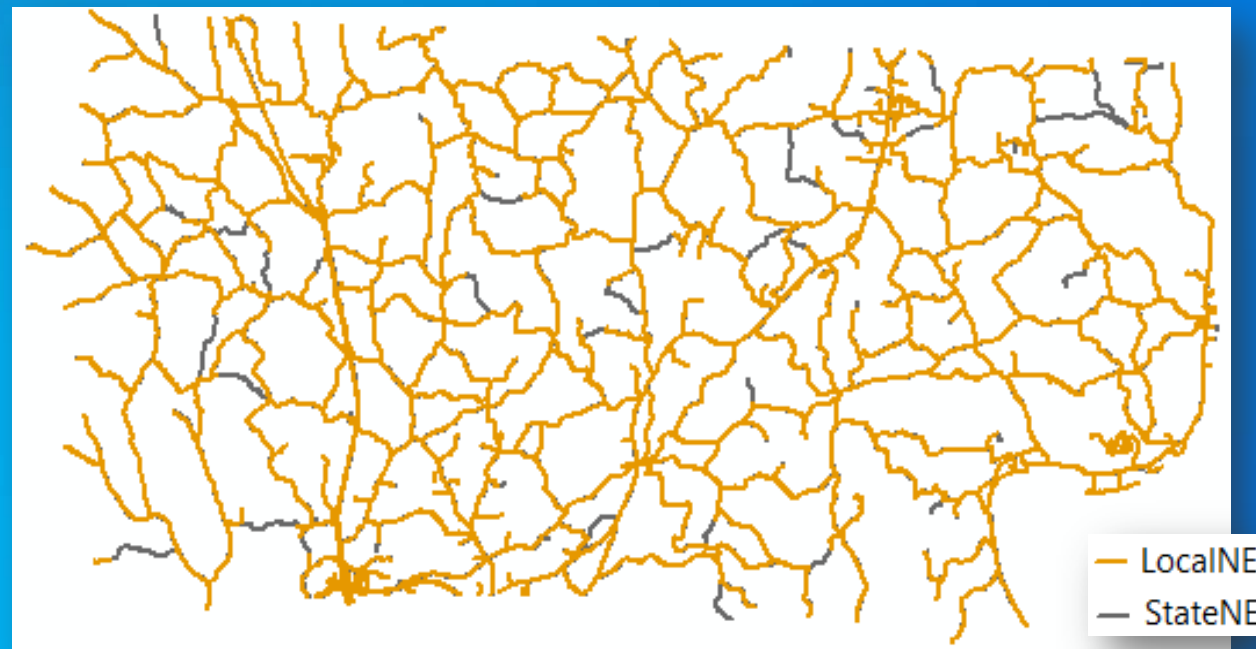
Data overview

Two road datasets
(northeast of Meigs
County, OH):

- **LocalNE – 1085 features**
- **StateNE – 1013 features**

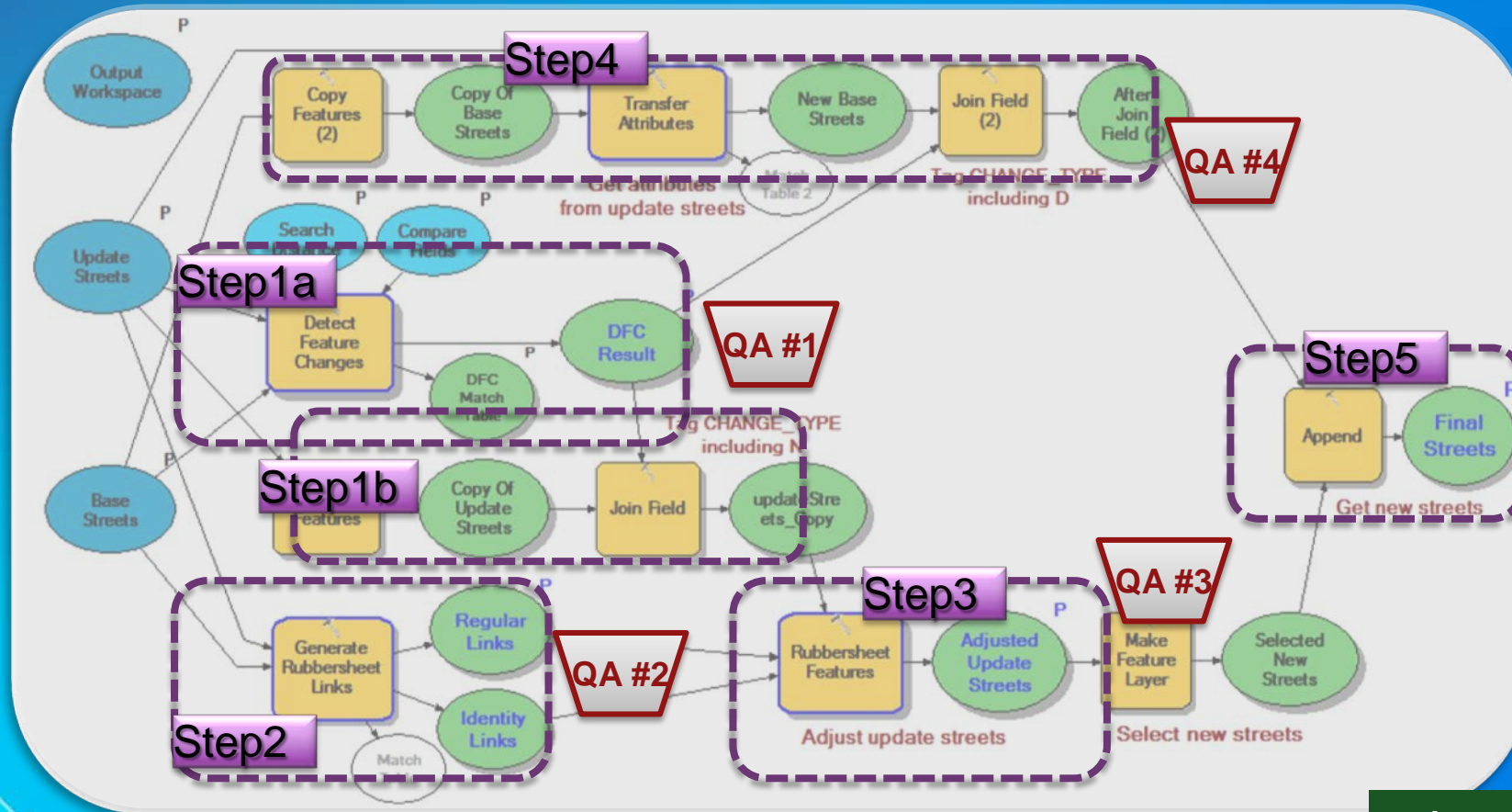
Both datasets:

- **Have common and uncommon features and attributes**
- **Are well preprocessed**



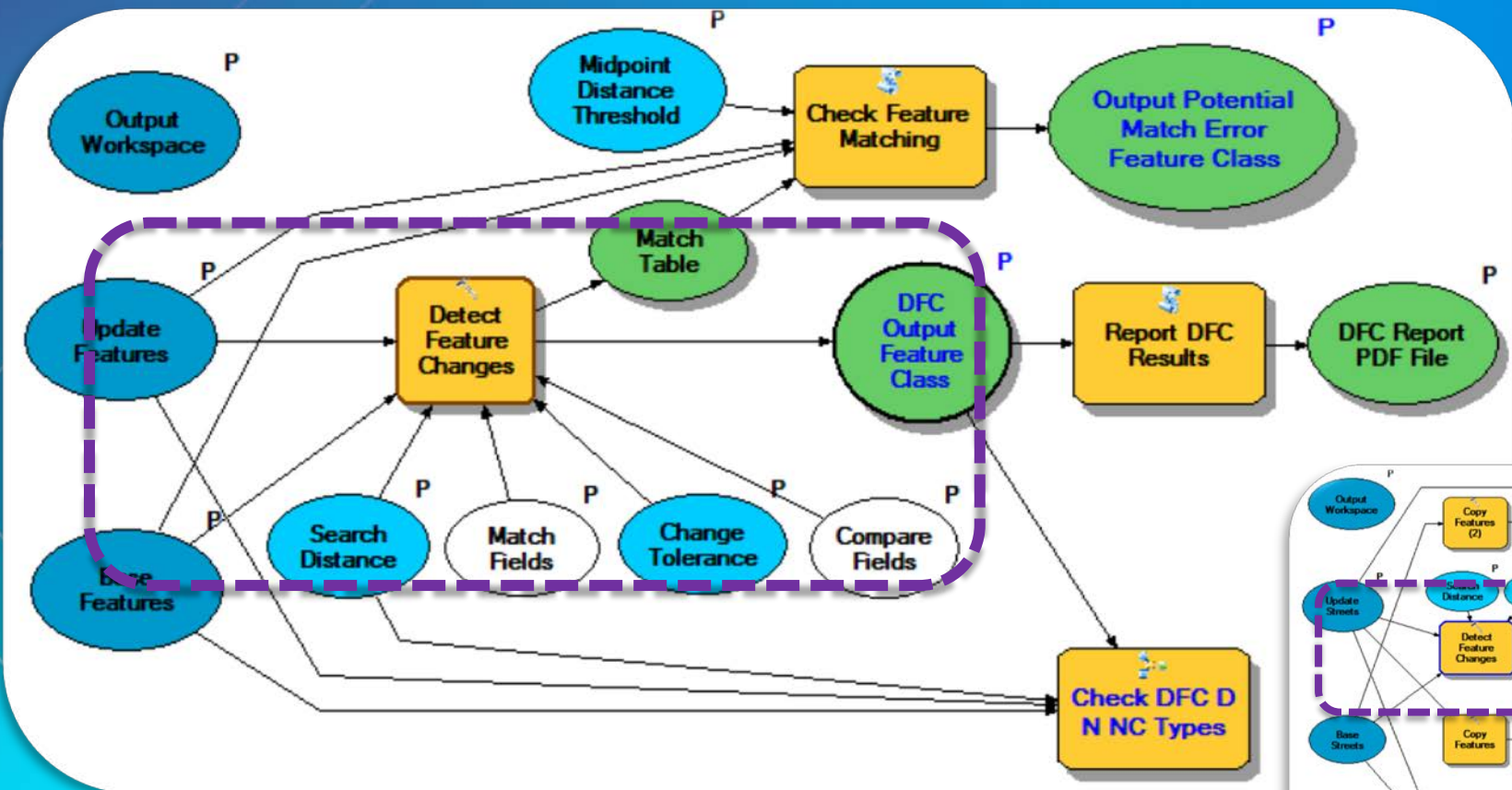
Breakdown of Demo 1 workflow into sub-workflows

Same goal and same strategy as Demo 1

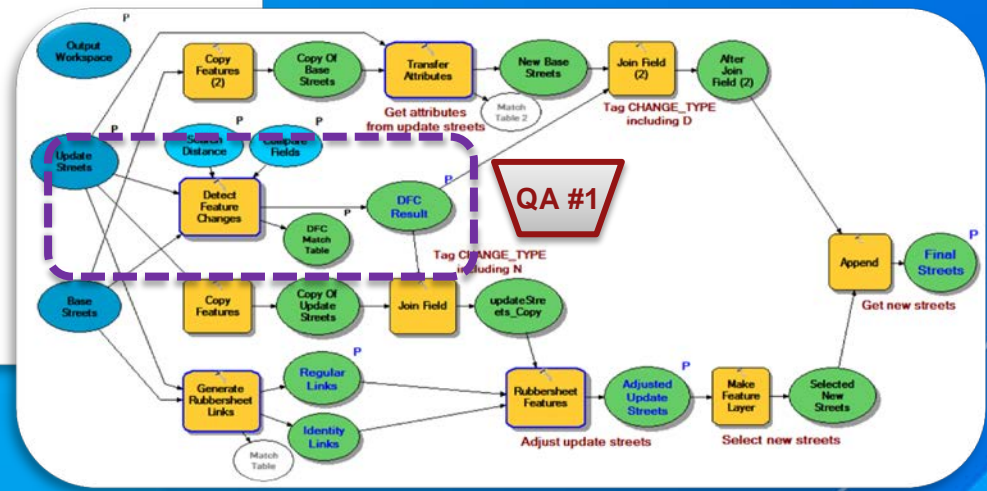


Let's get started ...

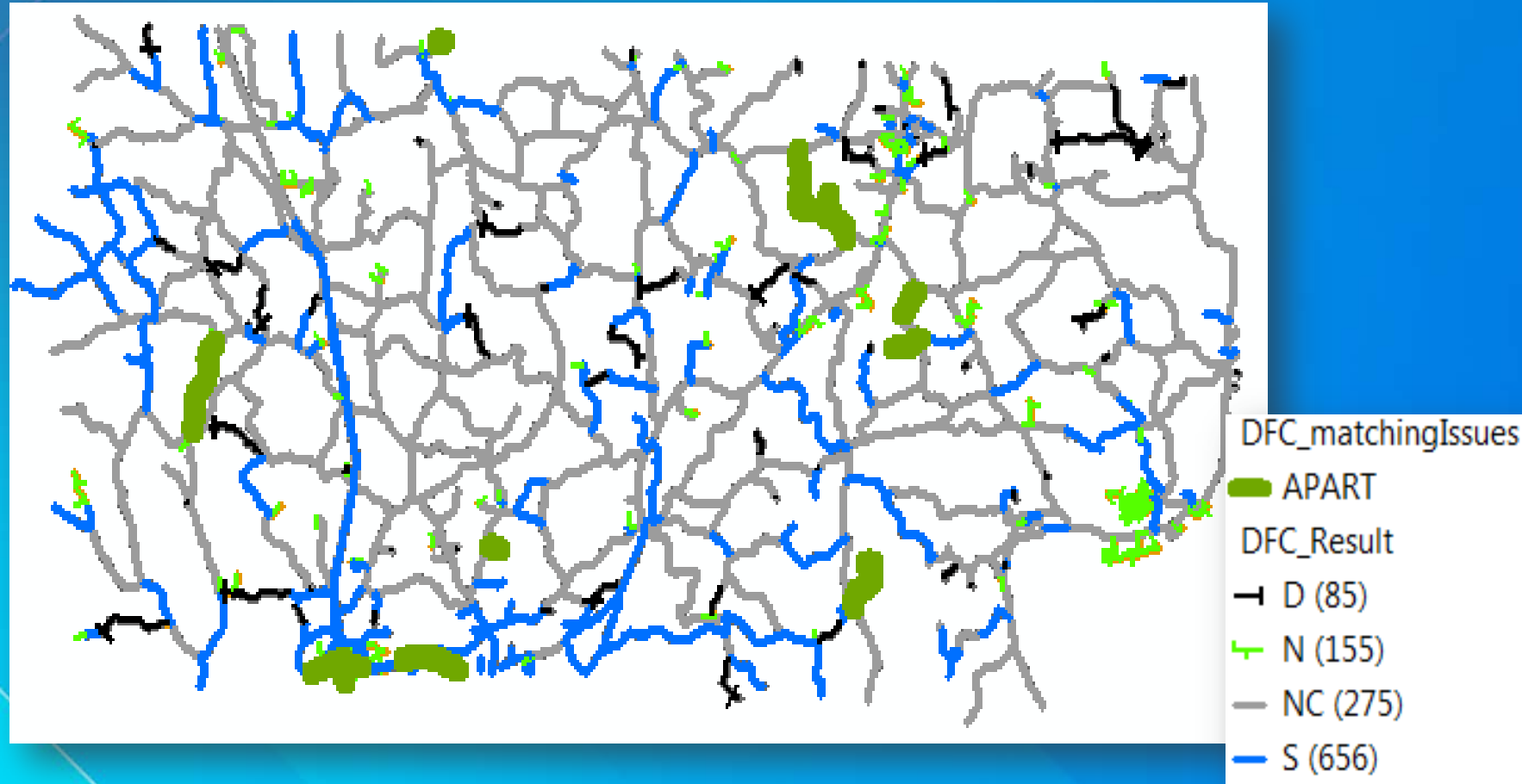
Step1a DFC and Evaluation



Step1a of the workflow with evaluation

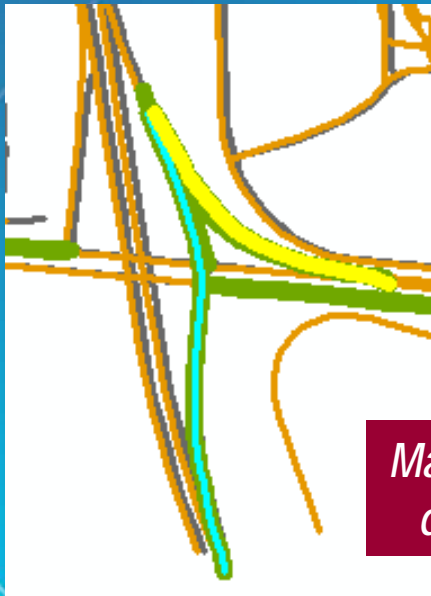


DFC result and potential match errors



QA potential match errors

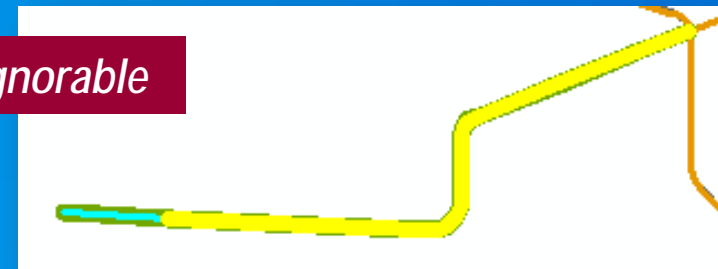
Total 16 CFM_GRP were flagged; 11 had match issues due to data complexity and dissimilarity; 5 were ignorable



Match issue due to data complexity

DFC_matchingIssues										
	OBJECTID *	Shape *	Shape_Length	SRC_FID	TGT_FID	FM_MN	FM_GRP	CFM_GRP	CFM_FLAG	CFM_DIST
▶	5	Polyline	130.569047	590	-1	2:1	25	2	APART	53.421149
	6	Polyline	742.933526	616	-1	2:1	25	2	APART	53.421149
	7	Polyline	748.586359	-1	550	2:1	25	2	APART	53.421149

Match issue ignorable



DFC_matchingIssues										
	OBJECTID *	Shape *	Shape_Length	SRC_FID	TGT_FID	FM_MN	FM_GRP	CFM_GRP	CFM_FLAG	CFM_DIST
	1	Polyline	344.068097	12	-1	3:1	13	1	APART	134.185317
	2	Polyline	22.322729	20	-1	3:1	13	1	APART	134.185317
	3	Polyline	188.622333	61	-1	3:1	13	1	APART	134.185317
	4	Polyline	360.206768	-1	62	3:1	13	1	APART	134.185317

QA DFC result – CHANGE_TYPE D and N

((CHANGE_TYPE = 'N') OR (CHANGE_TYPE = 'D')) AND((NEAR_DIST > 0) AND (NEAR_DIST < 10))

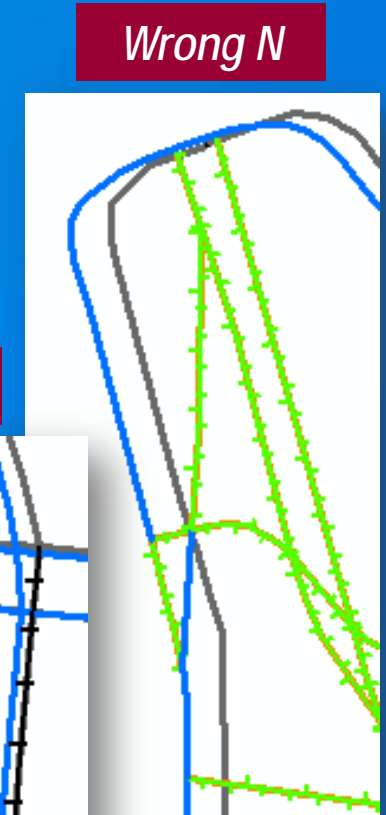
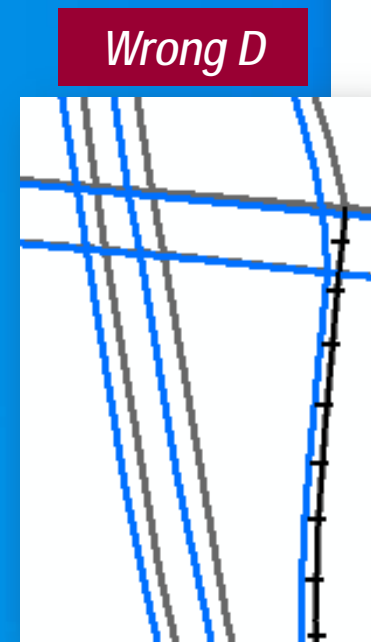
Inspect records with high potential for errors:

- 35 reviewed
- 11 wrong Ns or Ds flagged

DFC_Result_READY

UPDATE_FID	BASE_FID	CHANGE_TYPE	LEN_PCT	LEN_ABS	Shape_Length	NEAR_FID	NEAR_DIST	REV_FLAG
106	-1	N	-1	-1	30.065054	100	9.620227	wrongN
170	-1	N	-1	-1	132.907686	142	1.690797	wrongN
661	-1	N	-1	-1	13.923188	613	0.123593	wrongN
906	-1	N	-1	-1	36.27247	823	0.268347	wrongN
-1	1	D	-1	-1	312.204053	12	5.319732	wrongD
-1	10	D	-1	-1	23.34473	20	6.295733	wrongD
-1	66	D	-1	-1	204.019635	73	1.792909	wrongD
-1	96	D	-1	-1	0.718702	124	1.4274	wrongD
-1	142	D	-1	-1	132.812766	170	2.10387	wrongD
-1	613	D	-1	-1	13.919109	661	0.070317	wrongD
-1	823	D	-1	-1	36.257103	906	0.425543	wrongD
11	-1	N	-1	-1	167.678377	12	5.040875	isN
62	-1	N	-1	-1	122.483568	3	5.081806	isN
83	-1	N	-1	-1	13.434471	69	6.240292	isN

(35 out of 1171 Selected)



Feature matching accuracy estimates

Matched groups:

- Total: 896 groups
- Correct: 885 groups
- Incorrect: 11 groups

Accuracy = 885 / 896 = 98.77%

*Overall feature matching accuracy
(average of matched and unmatched)*

97.09%

Unmatched:

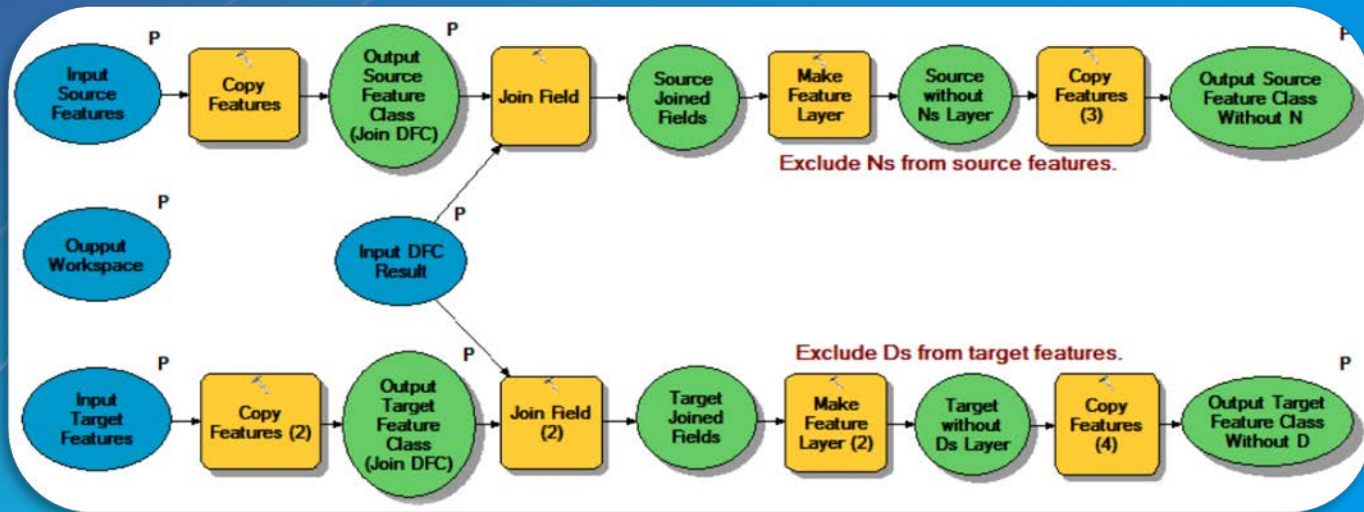
- Total: 240 (155 Ns + 85 Ds)
- Correct: 229 (151 Ns + 78 Ds)
- Incorrect: 11 (4 Ns + 7 Ds)

Accuracy = 229 / 240 = 95.42%

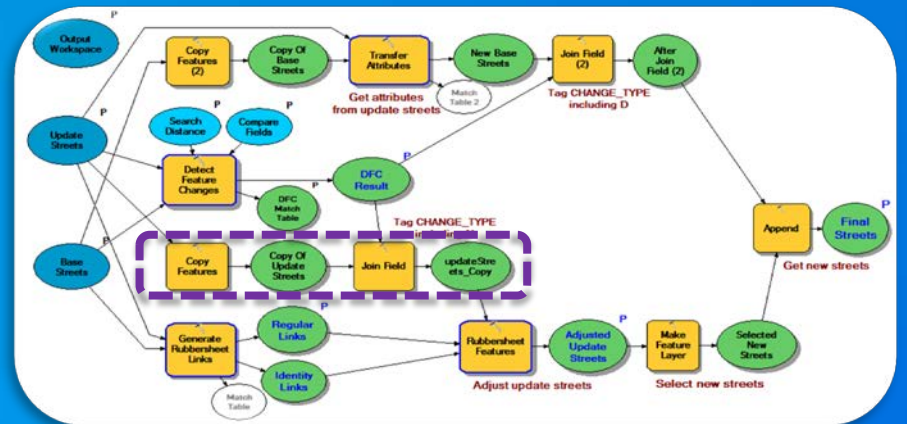
(biased by the total count)

*Ready to join with inputs to
tag Ns and Ds ...*

Step1b Extract Matched Features



Step1b of the workflow



Extract matched features
for GRL and TA processes

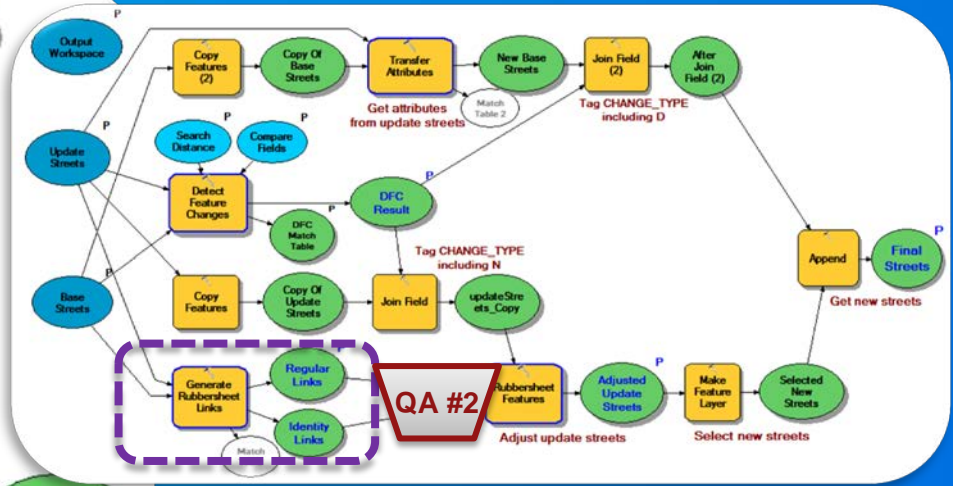
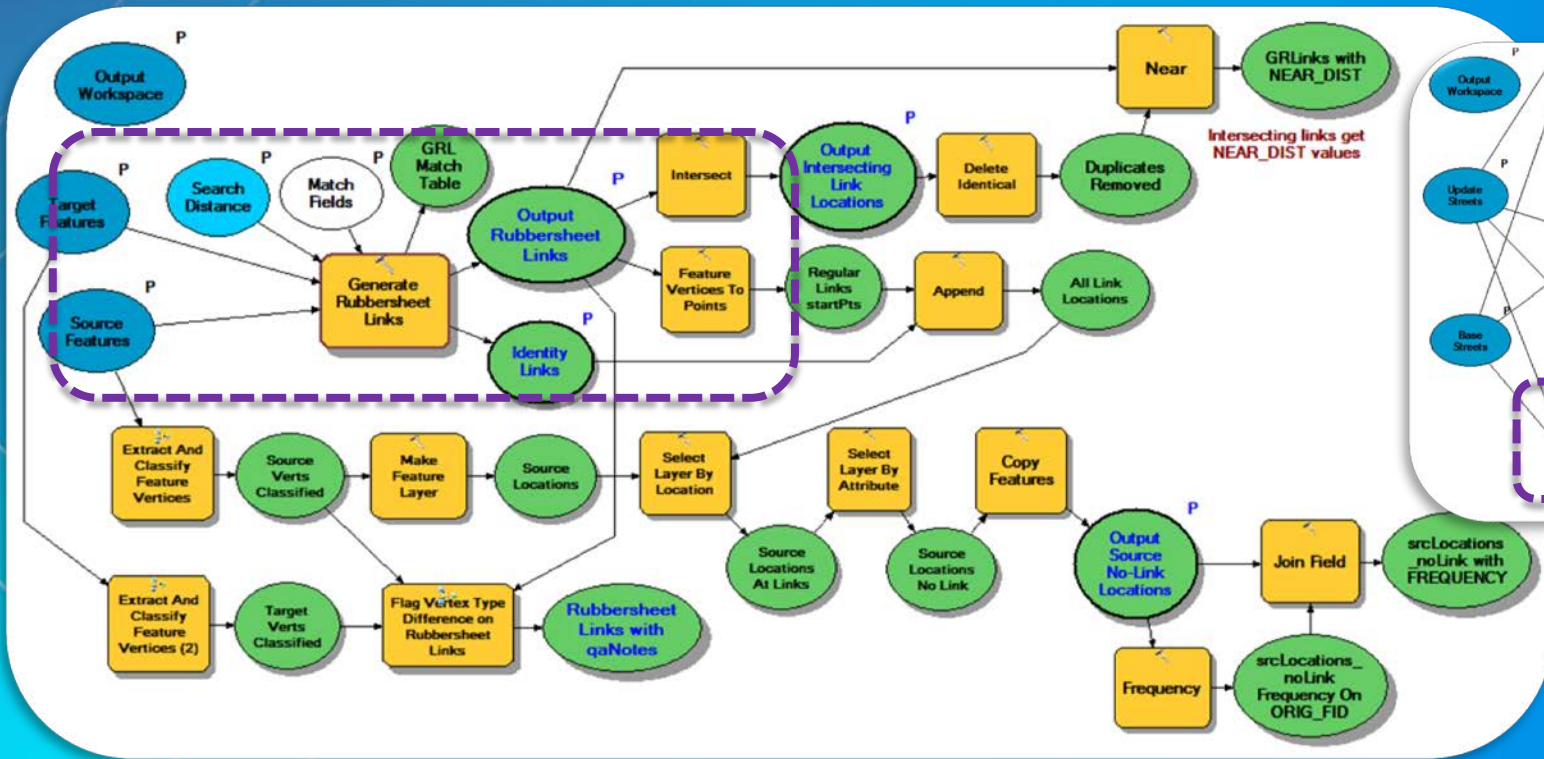
LocalNE: 934 non-N out of 1085
StateNE: 935 non-D out of 1013



Ready for GRL
process ...

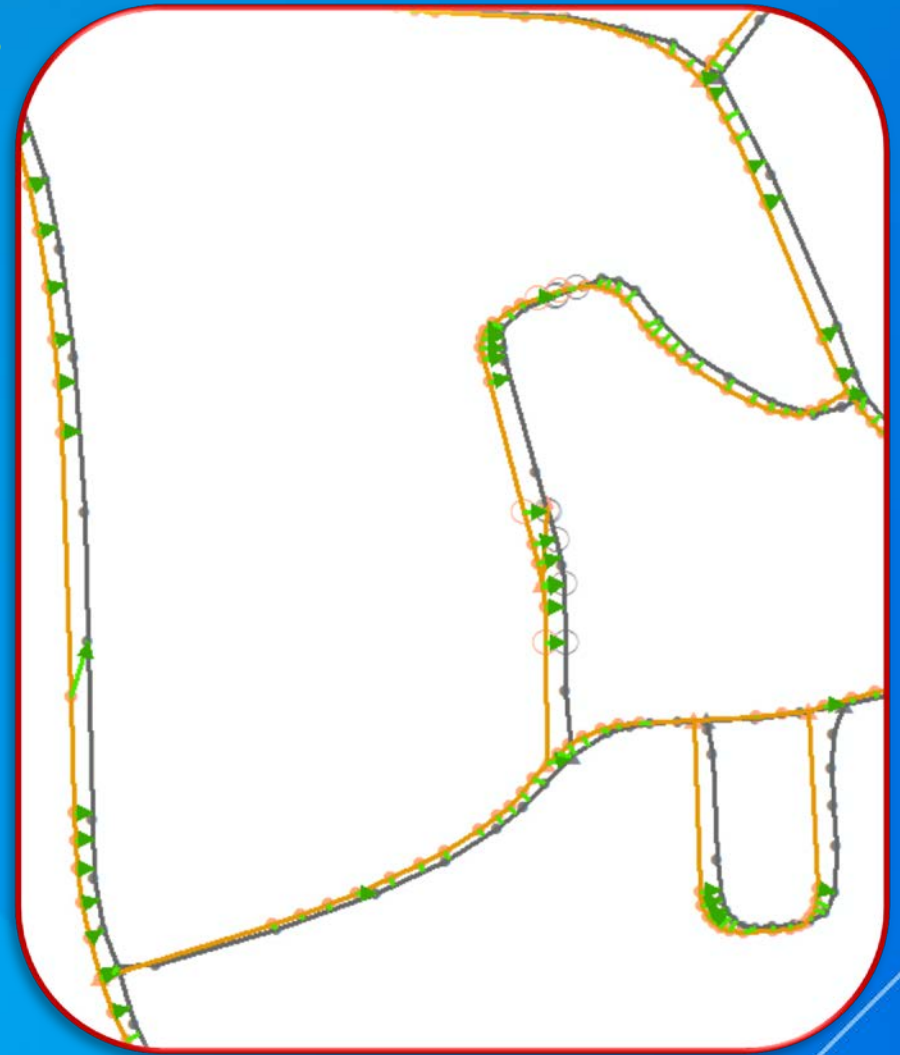
Step2 GRL and Evaluation

Step2 of the workflow with evaluation and QA



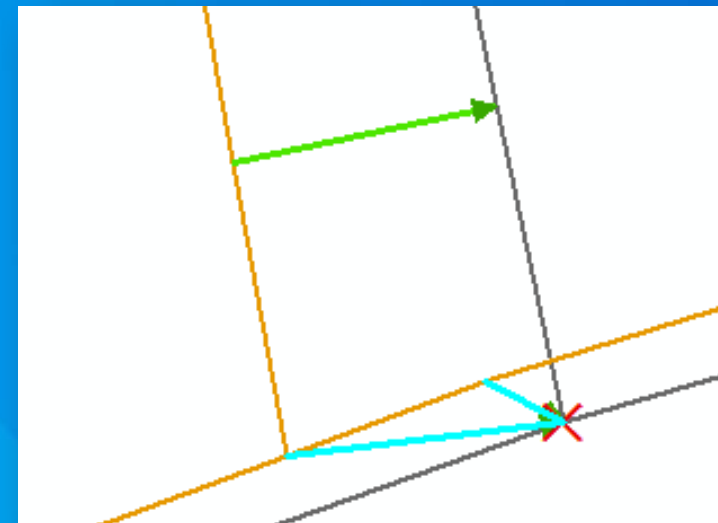
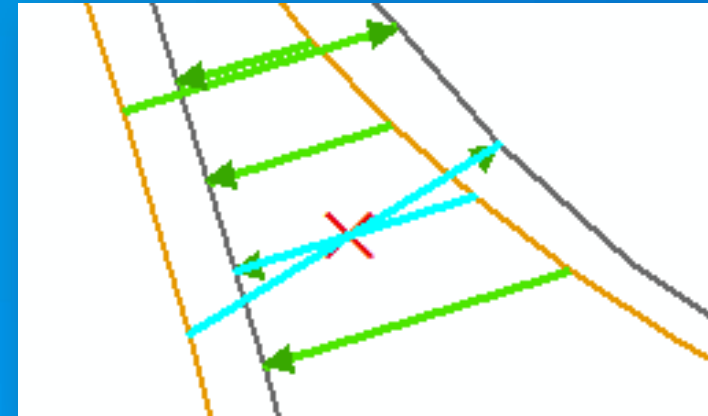
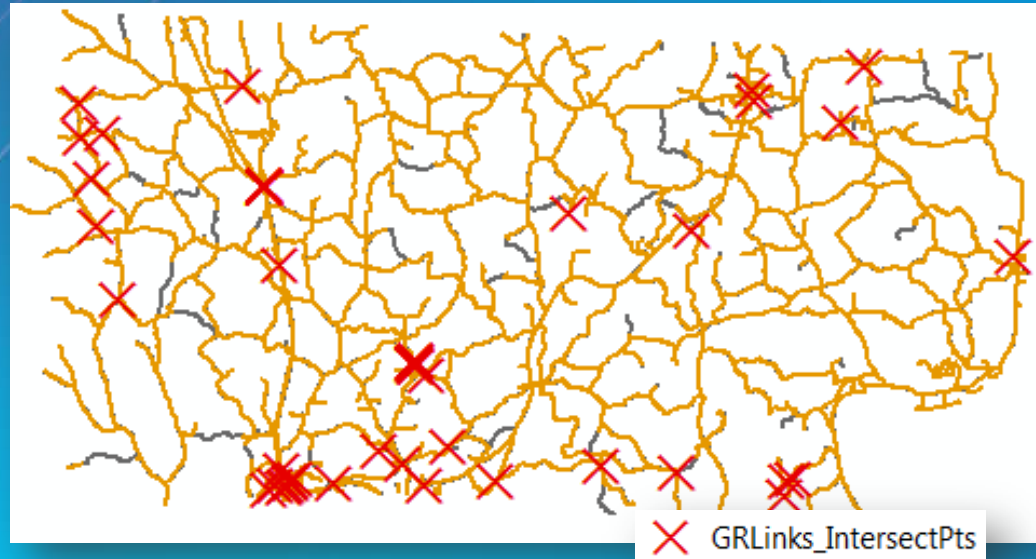
GRL result

Generated total 26198 regular links and 10227 identity links



GRL evaluation results – Intersecting links

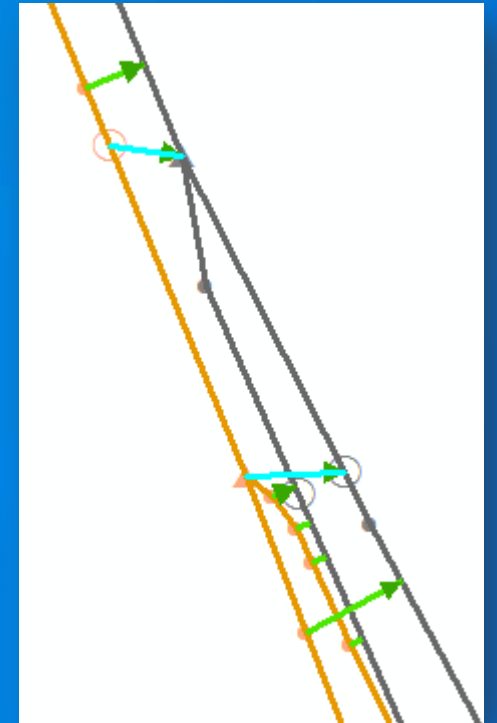
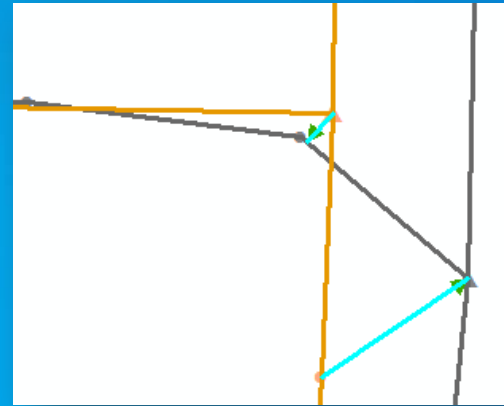
54 locations of intersecting links



OBJECTID	Shape *	SRC_FID	TGT_FID	Shape_Length	srcVxType	tgtVxType	qaNotes	NEAR_FID	NEAR_DIST
127	Polyline	52	58	42.244837	0	0	<Null>	1	0
1365	Polyline	53	56	29.023333	0	<Null>	src_tgt_VxType_diff	1	0
228	Polyline	-1	-1	10.842712	3	3	<Null>	3	0
1575	Polyline	81	-1	2.541051	0	3	src_tgt_VxType_diff	3	0
374	Polyline	-1	-1	4.855888	3	3	<Null>	5	0
393	Polyline	87	85	0.756891	0	3	src_tgt_VxType_diff	5	0
392	Polyline	86	-1	4.146372	0	3	src_tgt_VxType_diff	7	0
1534	Polyline	76	80	11.081514	3	3	<Null>	7	0
968	Polyline	-1	-1	32.61257	3	3	<Null>	9	0
969	Polyline	457	458	8.223584	0	3	src_tgt_VxType_diff	9	0

GRL evaluation results – linking different vertex types

(qaNotes = 'src_tgt_VxType_diff') AND((tgtVxType >=2) OR(srcVxType >=2)) AND NEAR_DIST = -1
79 of flagged links were more important

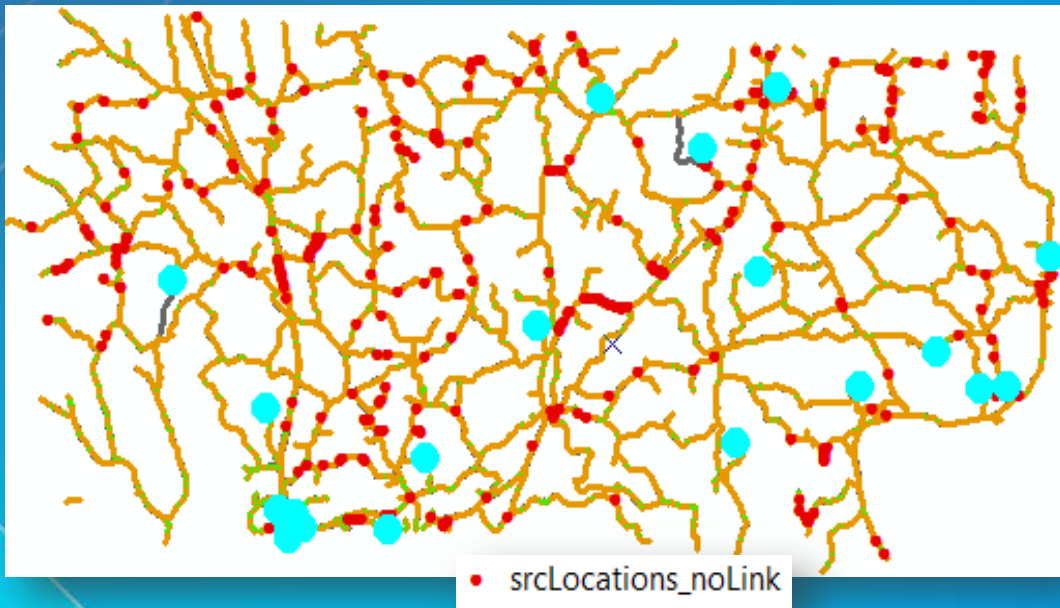


GRLinks										
	OBJECTID	Shape *	SRC_FID	TGT_FID	Shape_Length	srcVxType	tgtVxType	qaNotes	NEAR_FID	NEAR_DIST
	1049	Polyline	-1	-1	7.254691	3	4	src_tgt_VxType_diff	-1	-1
	120	Polyline	-1	-1	6.039016	2	3	src_tgt_VxType_diff	-1	-1
	1039	Polyline	-1	-1	7.148172	2	3	src_tgt_VxType_diff	-1	-1
	1497	Polyline	63	-1	9.38192	0	3	src_tgt_VxType_diff	-1	-1
	1508	Polyline	74	75	8.318095	0	3	src_tgt_VxType_diff	-1	-1
	4235	Polyline	221	-1	0.561213	0	3	src_tgt_VxType_diff	-1	-1
	10712	Polyline	420	-1	4.964278	1	3	src_tgt_VxType_diff	-1	-1
	15005	Polyline

Navigation: 0 (79 out of 26198 Selected)

GRL evaluation results – locations of missing links

22 of the 595 source locations of missing links were on nodes; all others are on in-line vertices.
20 ORIG_FID of frequency >5 locations were reviewed and confirmed non-critical.



srcLocations_noLink (595; 22 at nodes)

OBJECTID *	Shape *	ST_NAME	ORIG_FID	srcVxType
1	Point	SR 833 N TO US 33	3	4
20	Point	SR 833 N TO US 33	11	4
17	Point	SR 833 N TO US 33	3	2
19	Point	US 33	5	1
32	Point	US 33	53	1
57	Point	US 33	62	1
101	Point	MEIGS COUNTY FAIRGROUNDS	83	1
112	Point	FOX HILL	103	1
125	Point	RUSSELL	173	1
169	Point	LYONS DEN	184	1
207	Point	BLAKE HILL	258	1
211	Point	ODNR	292	1
217	Point	RIVERVIEW	295	1
220	Point	ANGELO	301	1
243	Point	CURTIS HOLLOW	371	1
250	Point	POOLER	385	1
303	Point	OSBORN	489	1
308	Point	MURRAY	504	1
332	Point	FIRST	524	1
482	Point	MERCER	691	1
522	Point	BOOTHE	801	1
550	Point	THIRD	855	1
2	Point	SR 833 N TO US 33	3	0
3	Point	SR 833 N TO US 33	3	0
4	Point	SR 833 N TO US 33	3	0
5	Point	SR 833 N TO US 33	3	0
6	Point	SR 833 N TO US 33	3	0
7	Point	SR 833 N TO US 33	3	0

QA regular links - summary

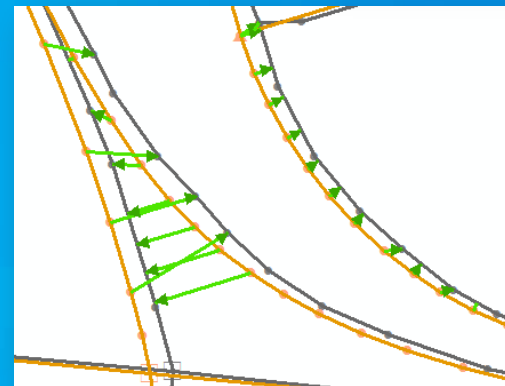
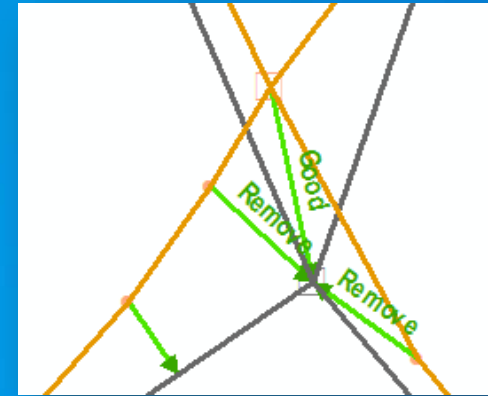
(qaNotes = 'src_tgt_VxType_diff') AND((tgtVxType >=2) OR(srcVxType >=2))

Total 241 (0.92%) of 26198 links were reviewed:

- 44 were modified
- 86 were to be removed
- 111 were ok

42 missing link locations were reviewed:

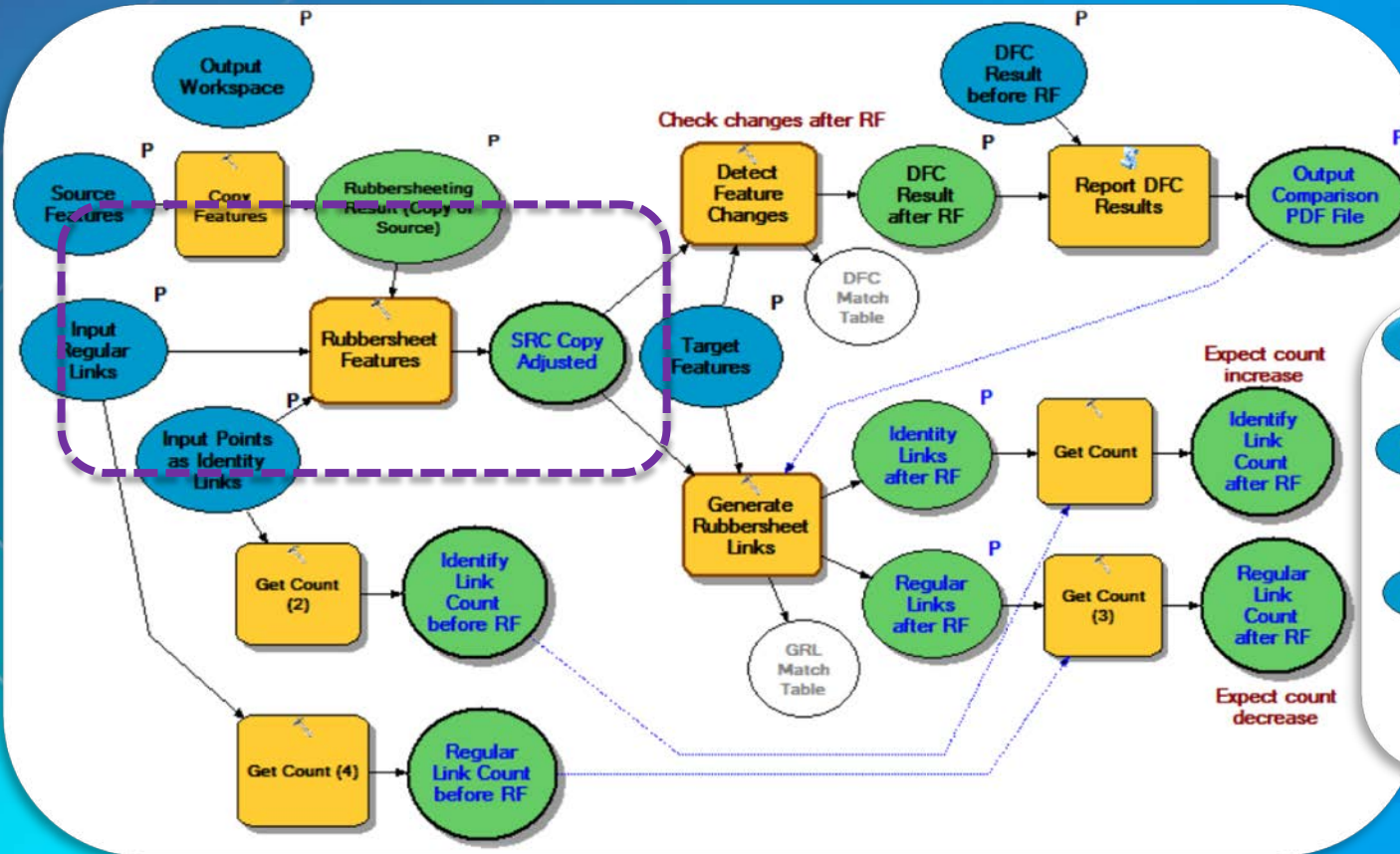
- 14 links were added
- Links at other locations were not critical



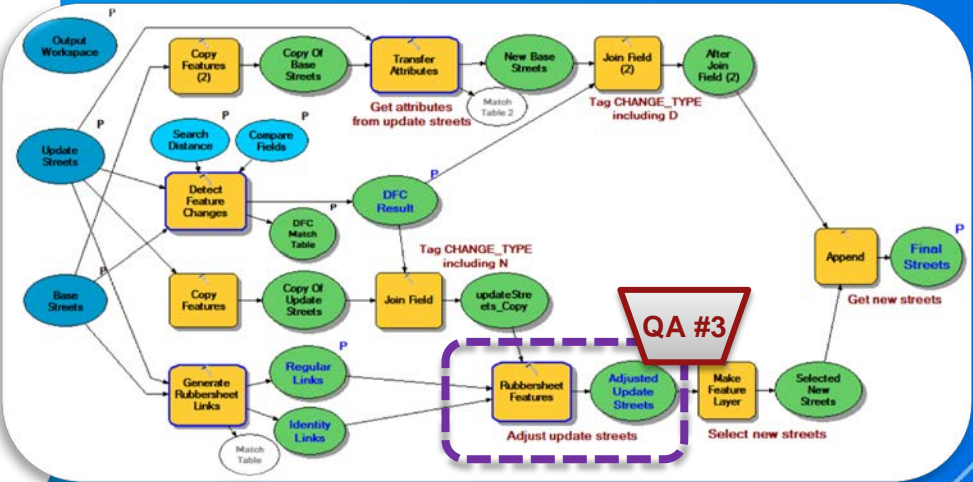
Ready for rubbersheeting ...

26126 regular links were selected by (REV_FLAG <> 'Delete') OR(REV_FLAG IS NULL) to participate

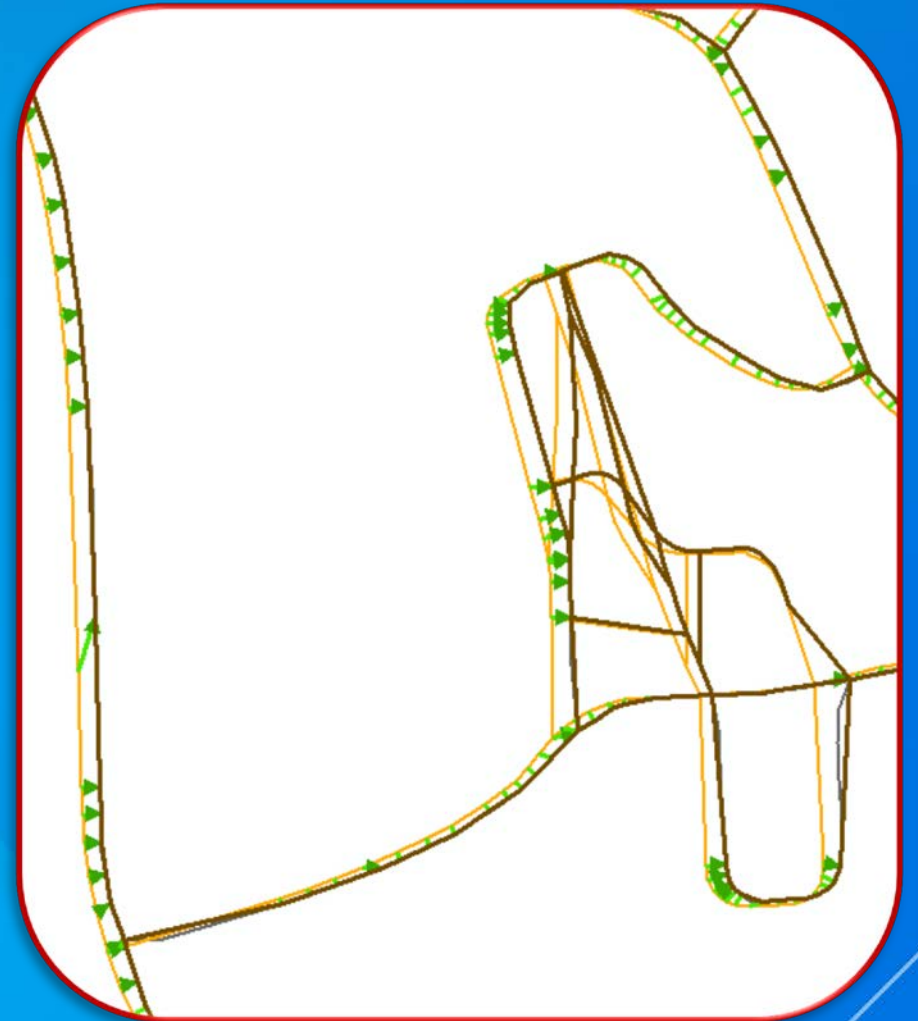
Step3 Rubbersheeting and Assessment



Step3 of the workflow with assessments

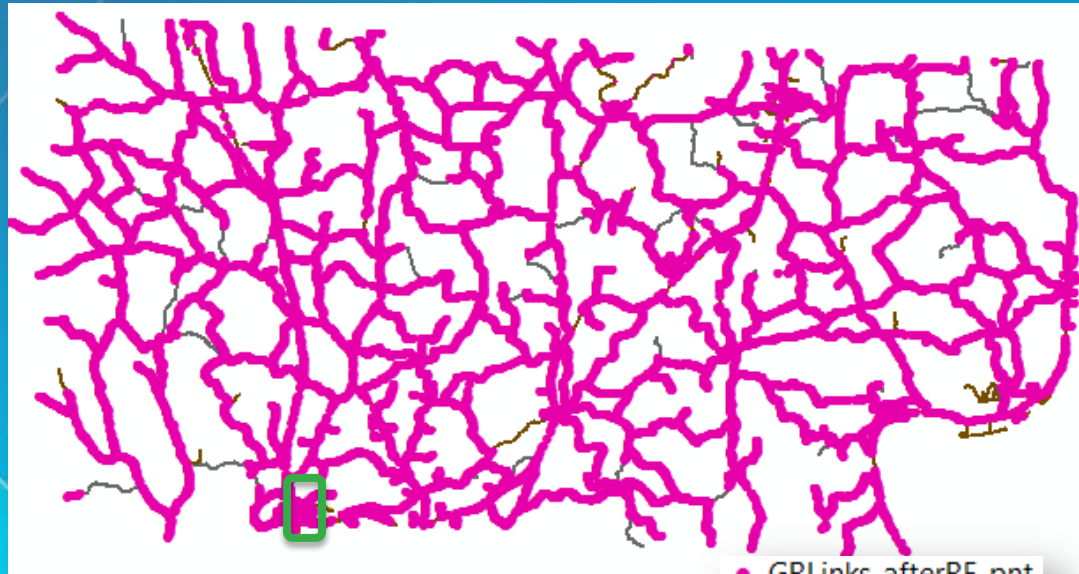


Rubbersheeting result

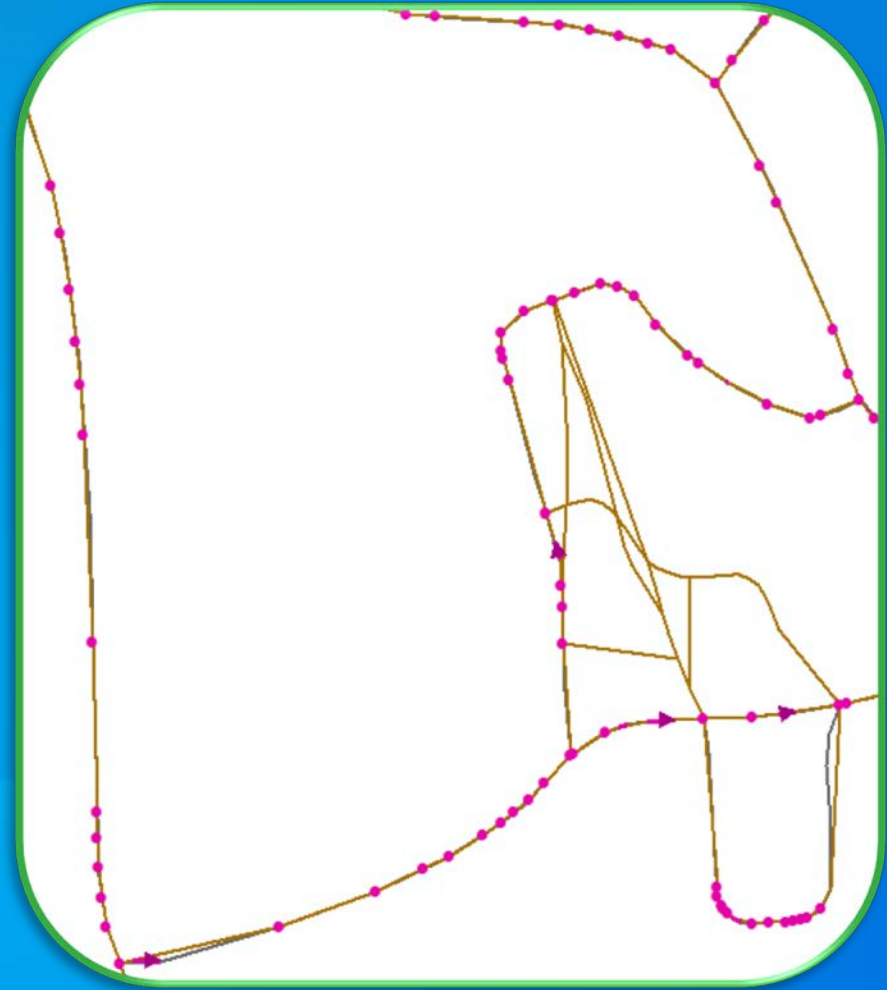


GRL result after rubbersheeting

Many regular links became identify links



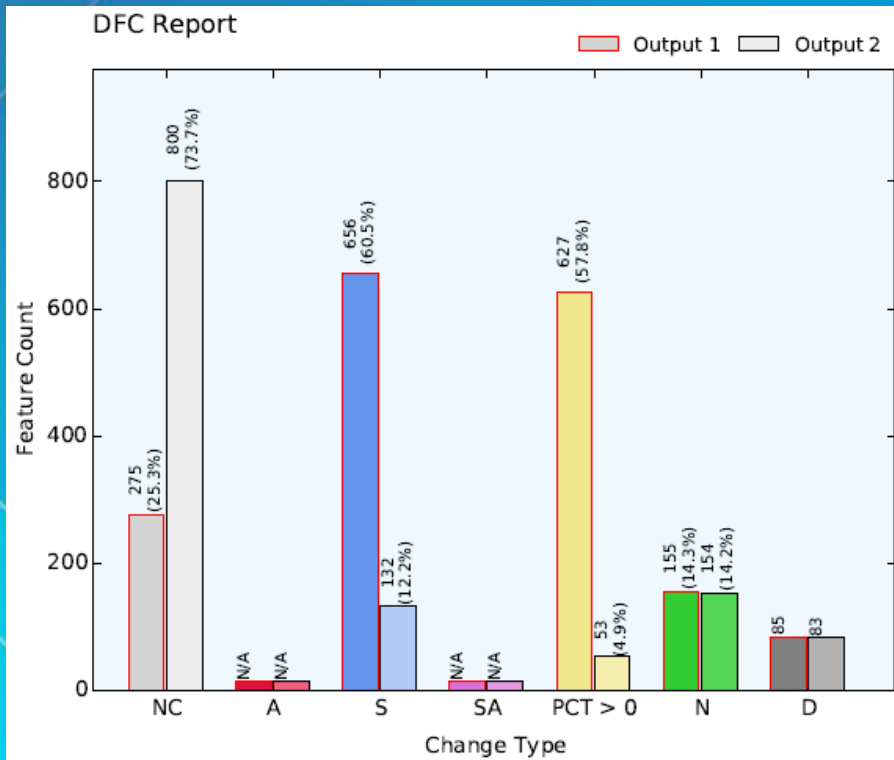
- GRLinks afterRF pnt
- GRLinks_afterRF



How good is the rubbersheeting result?

Three indicators showing spatial improvement

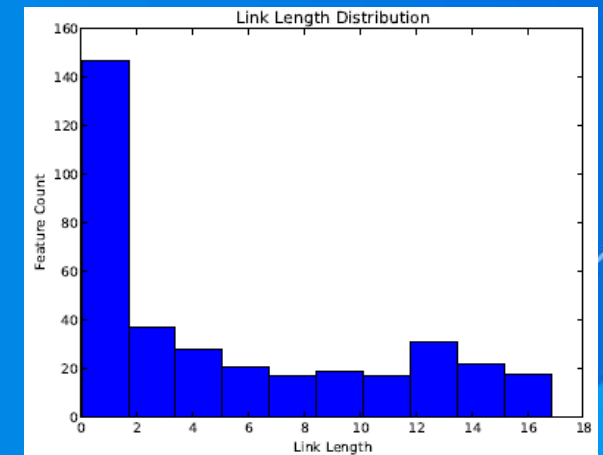
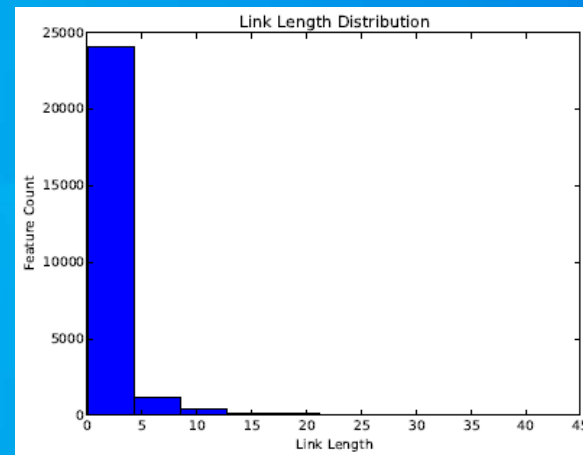
Less spatial differences



Improved location alignment

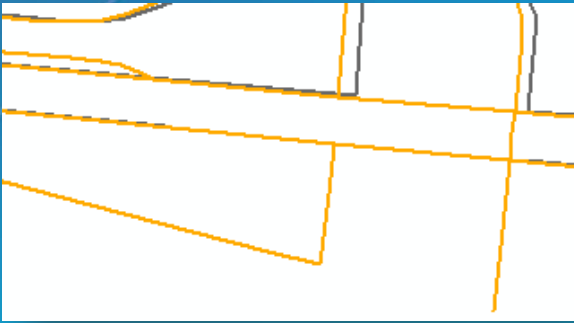
	Before RF	After RF
Regular links	26126	412
Identity links	10227	15456

*Link-length distributions before/after RF
- spatially closer to target
(Not on the same scale due to the big difference in values)*



QA #3 – Check rubbersheeting result

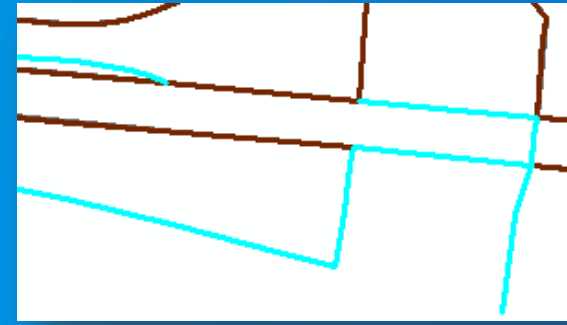
*Source (original)
and target*



Target features



*Source adjusted with
N features highlighted*

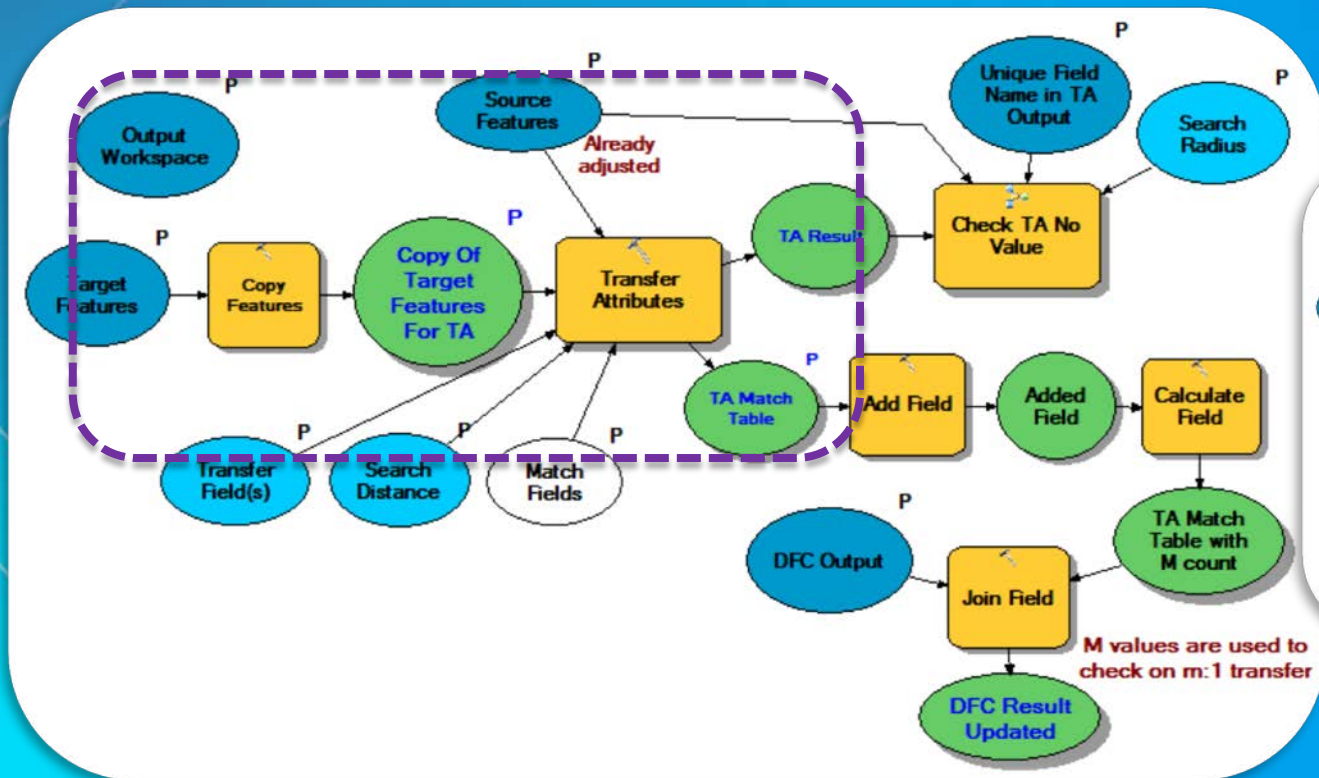


Ready to do TA ...

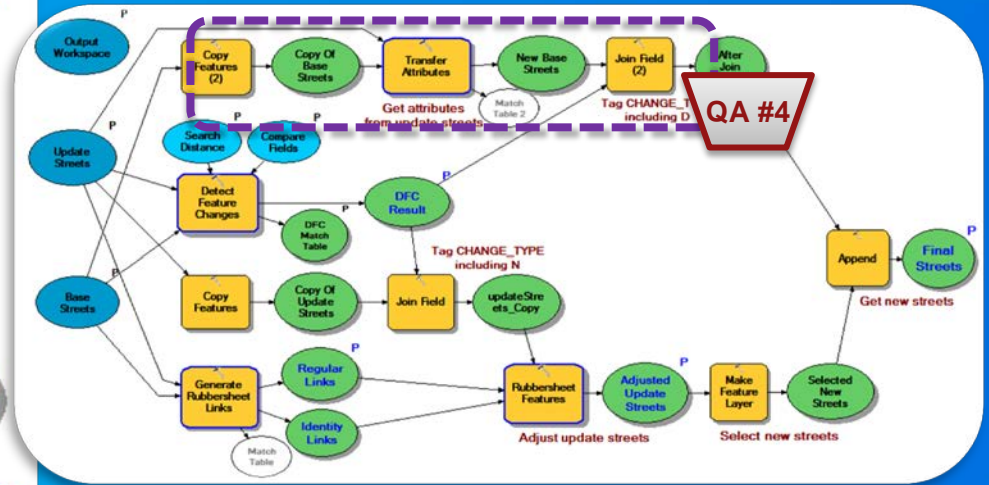
Transfer attributes from adjusted source to target

Excluding Ns from adjusted source; excluding Ds from joined target

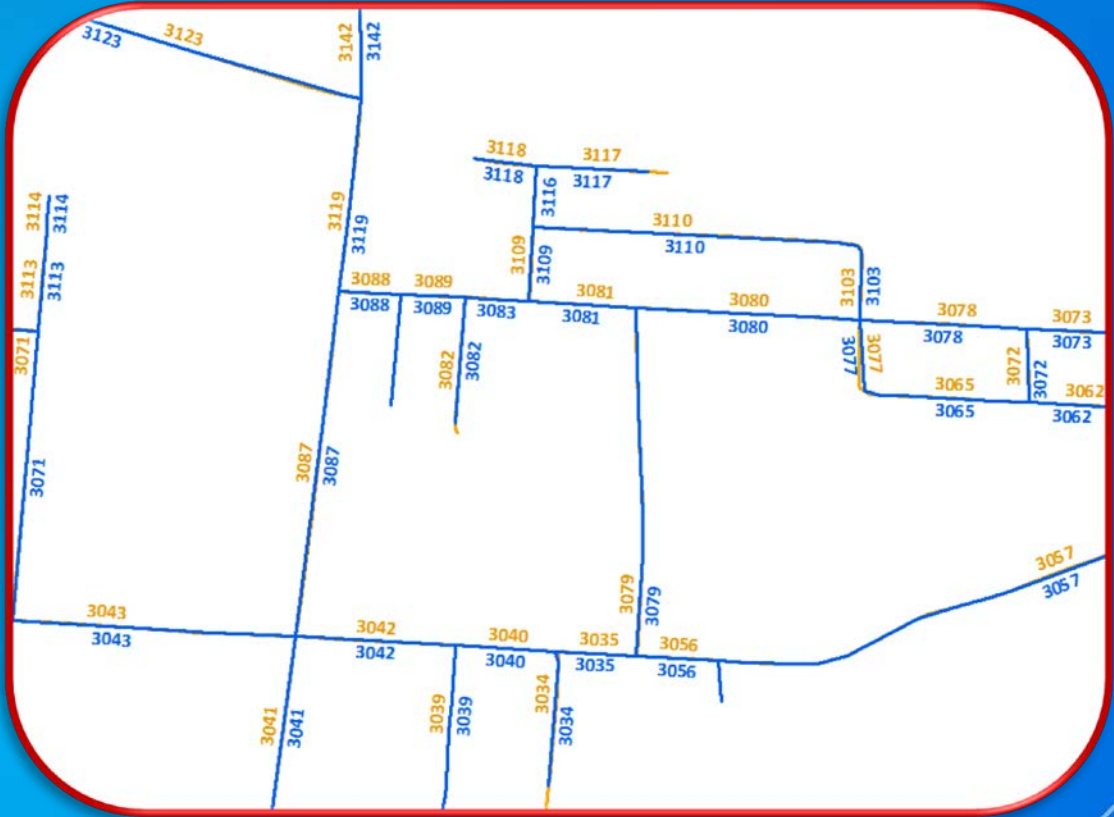
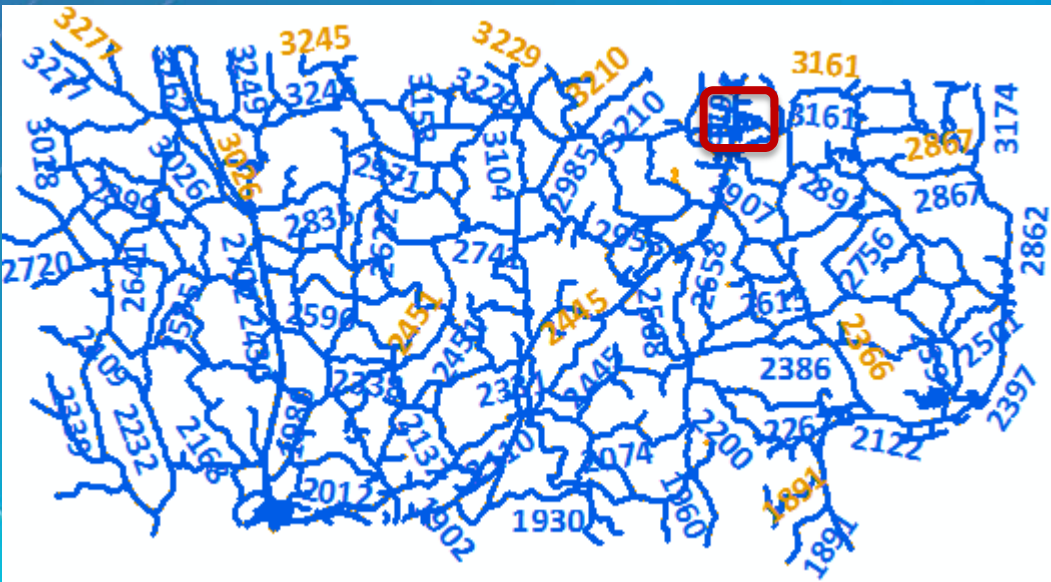
Step4 TA and Evaluation



Step4 of the workflow with evaluation and QA



Attribute transfer result



QA #4 – Check attribute transfer result

NEAR_DIST >=0; no-transfer features found nearby source features for potentially missed matches

32 records were reviewed:

- 18 were edited with UC2014_ID values

ID	CHANGE_TYPE	REV_FLAG	Shape_Length	RNAME	UC2014_ID	NEAR_FID	NEAR_DIST
25686	D	<Null>	59.713978	<Null>	<Null>	925	29.792733
30645	D	<Null>	56.086984	<Null>	<Null>	288	27.74351
30341	D	<Null>	55.831662	<Null>	<Null>	634	27.658837
30510	D	<Null>	95.916216	<Null>	<Null>	936	26.24684
29763	D	<Null>	37.817835	<Null>	<Null>	58	23.862568
30473	D	<Null>	44.24111	<Null>	<Null>	222	22.120555
30563	D	<Null>	432.374857	<Null>	<Null>	69	21.811976
30471	D	<Null>	74.115629	<Null>	<Null>	462	21.721167
30118	D	<Null>	42.415217	<Null>	<Null>	679	21.207608
30561	D	<Null>	49.517875	<Null>	<Null>	66	20.781753
30426	D	<Null>	40.081914	<Null>	<Null>	250	19.575952
29775	D	<Null>	39.455303	<Null>	<Null>	934	19.526662
30362	D	<Null>	38.453231	<Null>	<Null>	963	19.201207
29968	D	<Null>	40.817817	<Null>	<Null>	248	17.860293
30644	D	<Null>	83.30584	<Null>	<Null>	432	16.330416
30266	D	<Null>	41.712116	<Null>	<Null>	368	16.252561
30435	D	<Null>	30.597128	<Null>	<Null>	650	15.282429
30069	D	<Null>	30.119045	<Null>	<Null>	171	15.232608
30620	D	<Null>	30.719093	<Null>	<Null>	515	15.014051
30469	D	<Null>	318.801312	<Null>	<Null>	606	14.967127
30008	D	<Null>	153.620314	<Null>	<Null>	261	11.071826
29744	D	wrongD	23.34473	<Null>	<Null>	20	6.295733
29743	D	wrongD	312.204053	<Null>	<Null>	12	5.319732
30772	D	isD	26.47975	<Null>	<Null>	58	4.712168
30862	D	isD	5.045483	<Null>	<Null>	262	2.519857
29875	D	isD	4.549253	<Null>	<Null>	260	2.273373
30404	D	wrongD	132.812766	<Null>	<Null>	170	2.10387
30750	D	wrongD	204.019635	<Null>	<Null>	73	1.792909
29373	D	wrongD	0.718702	<Null>	<Null>	124	1.4274
30790	D	wrongD	36.257103	<Null>	<Null>	906	0.425543
30305	D	wrongD	13.919109	<Null>	<Null>	661	0.070317
30240	D	isD	0.014701	<Null>	<Null>	363	0.002076
30234	D	<Null>	1343.361864	<Null>	<Null>	-1	-1
30404	D	<Null>	2051.673374	<Null>	<Null>	-1	-1
30462	D	<Null>	266.543937	<Null>	<Null>	-1	-1
30169	D	<Null>	860.788377	<Null>	<Null>	-1	-1
30269	D	<Null>	498.717165	<Null>	<Null>	-1	-1

CHANGE_TYPE	REV_FLAG	Shape_Length	RNAME	UC2014_ID	NEAR_FID	NEAR_DIST
D	<Null>	59.713978	<Null>	3048	925	29.792733
D	wrongD	36.257103	<Null>	3027	906	0.425543
D	<Null>	42.415217	<Null>	2688	679	21.207608
D	wrongD	13.919109	<Null>	2661	661	0.070317
D	<Null>	30.597128	<Null>	2648	650	15.282429
D	<Null>	55.831662	<Null>	2627	634	27.658837
D	<Null>	30.719093	<Null>	2456	515	15.014051
D	<Null>	56.086984	<Null>	2132	288	27.74351
D	<Null>	40.081914	<Null>	2067	250	19.575952
D	<Null>	40.817817	<Null>	2064	248	17.860293
D	<Null>	44.24111	<Null>	2022	222	22.120555
D	<Null>	30.119045	<Null>	1937	171	15.232608
D	wrongD	132.812766	<Null>	1935	170	2.10387
D	wrongD	0.718702	<Null>	1874	124	1.4274
D	wrongD	204.019635	<Null>	1820	73	1.792909
D	<Null>	49.517875	<Null>	1912	66	20.781753
D	wrongD	23.34473	<Null>	1764	20	6.295733
D	wrongD	312.204053	<Null>	1755	12	5.319732
D	isD	26.47975	<Null>	<Null>	58	4.712168
D	<Null>	432.374857	<Null>	<Null>	69	21.811976
D	<Null>	37.817835	<Null>	<Null>	58	23.862568
D	isD	5.045483	<Null>	<Null>	262	2.519857
D	<Null>	153.620314	<Null>	<Null>	261	11.071826
D	isD	4.549253	<Null>	<Null>	260	2.273373
D	<Null>	41.712116	<Null>	<Null>	368	16.252561
D	isD	0.014701	<Null>	<Null>	363	0.002076
D	<Null>	74.115629	<Null>	<Null>	462	21.721167
D	<Null>	83.30584	<Null>	<Null>	432	16.330416
D	<Null>	318.801312	<Null>	<Null>	606	14.967127
D	<Null>	38.453231	<Null>	<Null>	963	19.201207
D	<Null>	95.916216	<Null>	<Null>	936	26.24684
D	<Null>	39.455303	<Null>	<Null>	934	19.526662

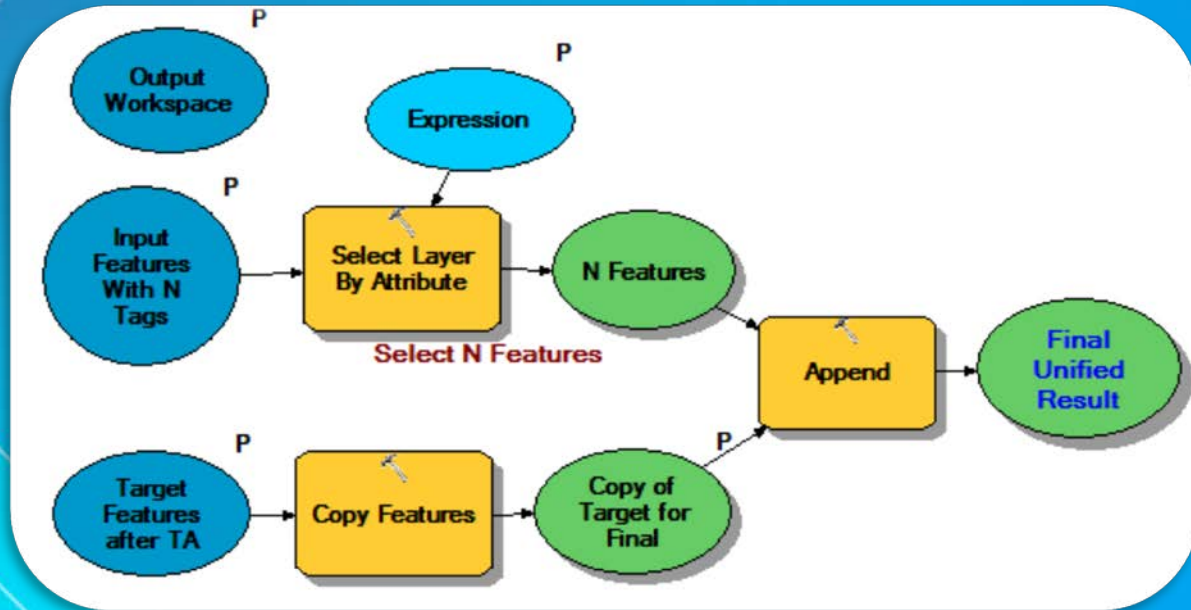


Almost there ...

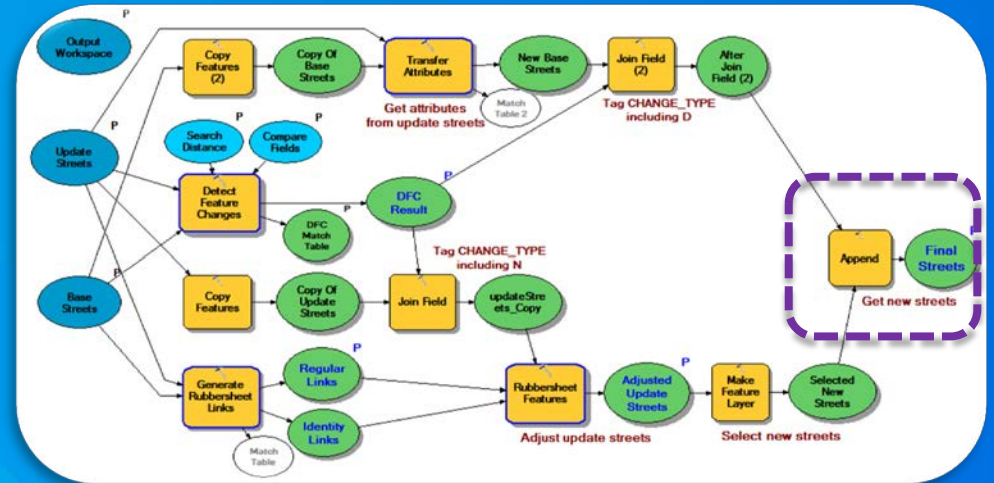
Select adjusted N features; append them to target

(CHANGE_TYPE = 'N') AND((REV_FLAG <> 'wrongN') OR REV_FLAG IS NULL)

Step5 Append N For Final



Final step of the full workflow

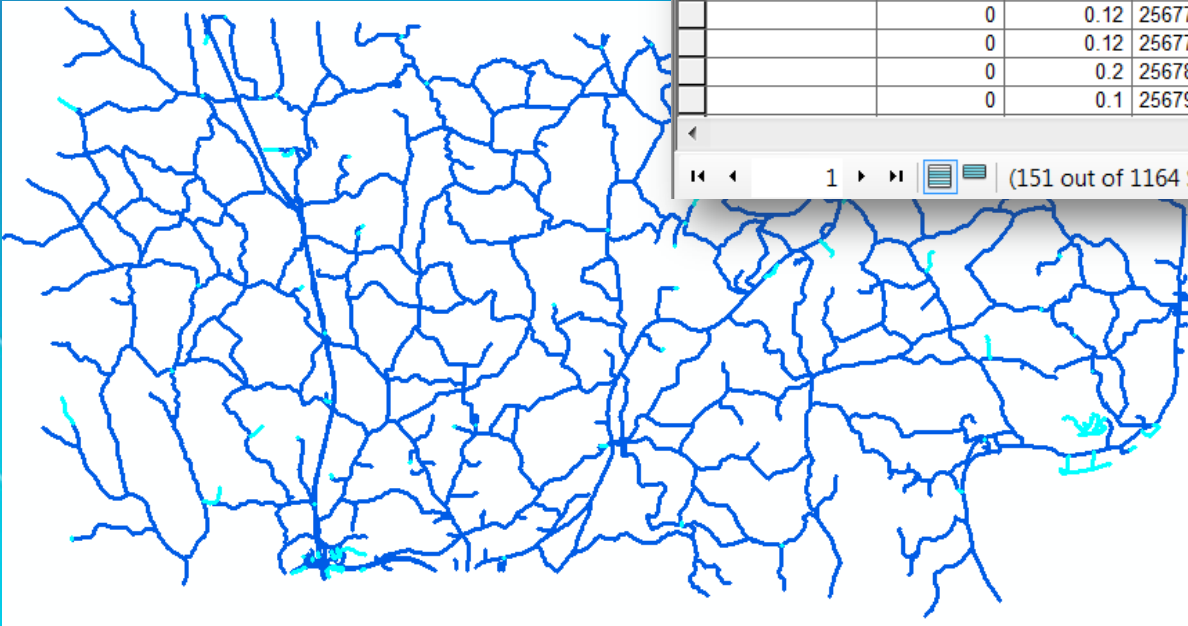


Appended N features in final result

tgt_CopyForTA_FINAL

ROAD_STATU	LOG_BEG	LOC_END	ID	CHANGE_TYPE	REV_FLAG	Shape_Lengt	RNAME	UC2014_ID	NEAR_FID	NEAR_DIST
<Null>	<Null>	<Null>	<Null>	N	<Null>	58.002288	GUTHRIE	3196	<Null>	<Null>
<Null>	<Null>	<Null>	<Null>	N	isN	23.660369	BUCK LAKE	3200	<Null>	<Null>
<Null>	<Null>	<Null>	<Null>	N	isN	0.39117	MUDSOCK	3207	<Null>	<Null>
<Null>	<Null>	<Null>	<Null>	N	<Null>	45.454298	BEARWALLO	3221	<Null>	<Null>
<Null>	<Null>	<Null>	<Null>	N	<Null>	210.402302	GRUESER	3237	<Null>	<Null>
<Null>	<Null>	<Null>	<Null>	N	isN	0.013617	HELWIG RID	3261	<Null>	<Null>
<Null>	<Null>	<Null>	<Null>	N	isN	6.392112	HELWIG RID	3263	<Null>	<Null>
	0	0.08	25675	S	<Null>	70.444146	ELM	3109	<Null>	<Null>
	0	0.08	25675	NC	<Null>	58.467636	ELM	3116	<Null>	<Null>
	0	0.08	25676	D	<Null>	107.125884	<Null>	<Null>	-1	-1
	0	0.12	25677	S	<Null>	107.147296	THIRD	3117	<Null>	<Null>
	0	0.12	25677	S	<Null>	59.466139	THIRD	3118	<Null>	<Null>
	0	0.2	25678	S	<Null>	317.388112	SECOND	3110	<Null>	<Null>
	0	0.1	25679	S	<Null>	68.438521	POPLAR	3077	<Null>	<Null>

1 (151 out of 1164 Selected)



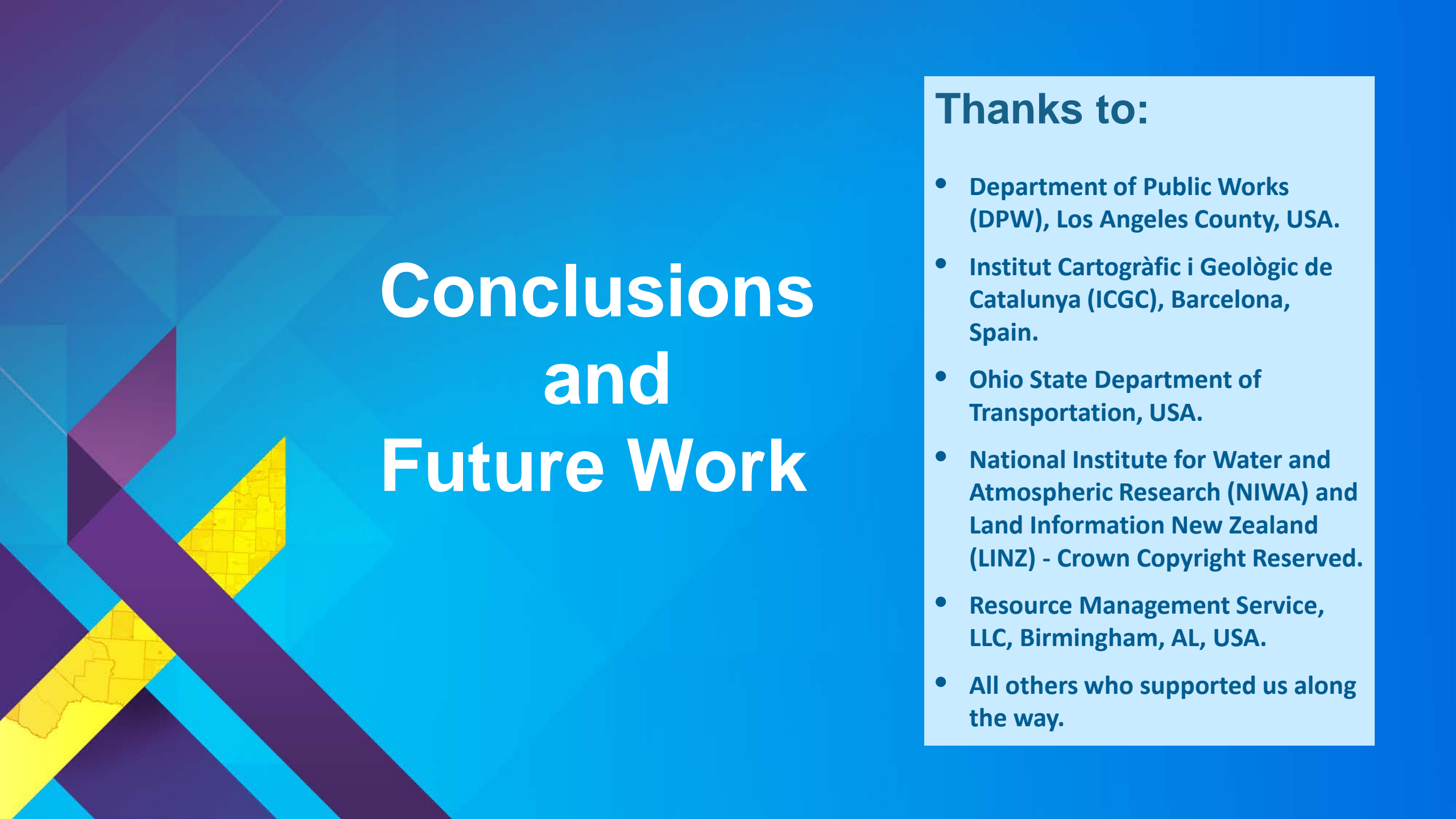
Unification of overlapping datasets completed!

Automated processing

	Processing Time
Step 1 (a, b)	1 min 3 sec
Step 2	1 min 14 sec
Step 3	1 min
Steps 4, 5	18 sec
Total	3 min 35 sec

*Interactive processing
(not counting final review)*

	QA #1 (CFM_GR P and DN)	QA #2 (links)	QA #3	QA #4 (attribute transfer)	QA Total	Time (2-3 review counts per minute)
Review Count (locations or feature groups)	51	283	x	32	366	~ 2-3 hrs.
Edit Count (field values)	46	255	x	18	319	

The background features a blue gradient with geometric shapes in purple, yellow, and cyan. A semi-transparent map of a region is overlaid on the left side.

Conclusions and Future Work

Thanks to:

- Department of Public Works (DPW), Los Angeles County, USA.
- Institut Cartogràfic i Geològic de Catalunya (ICGC), Barcelona, Spain.
- Ohio State Department of Transportation, USA.
- National Institute for Water and Atmospheric Research (NIWA) and Land Information New Zealand (LINZ) - Crown Copyright Reserved.
- Resource Management Service, LLC, Birmingham, AL, USA.
- All others who supported us along the way.

Conflation can be done more efficiently now

It takes a workflow:

- **Use the best practice in preprocessing.**
- **Highly accurate results and rich information are produced automatically.**
- **Small amount of interactive review and editing are necessary; time is worth-spending.**

Consider conflation a higher priority

Study the tools and workflows; understand the results

- **Start with small test areas**

Customize the workflows for your organizations

- **Improve data quality and usability**
- **Bring new life and value to your data**

Work with broader communities

- **Data sharing and collaboration**
- **Seamless analysis and mapping**

*Please send us your
feedbacks 😊*

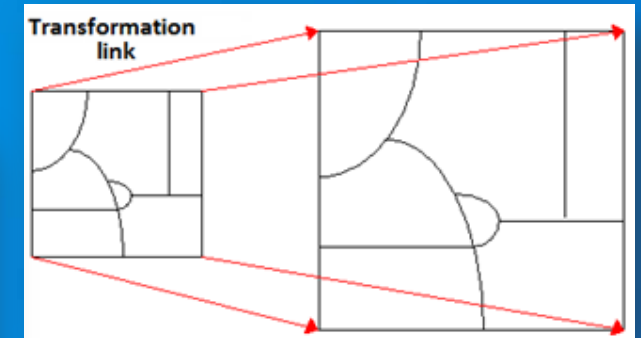
Our future work

New tools and enhancements

- **New option in DFC tool: Compare line directions**
- **New Gp tools: Transform Features, Align Features**
- **Better feature matching**

Formalization of workflows

- **Common scenarios oriented**
- **Integrated review and editing**
- **Other feature types**
- **Contextual conflation (spatially related features)**



Please send us your use cases and requirements ☺

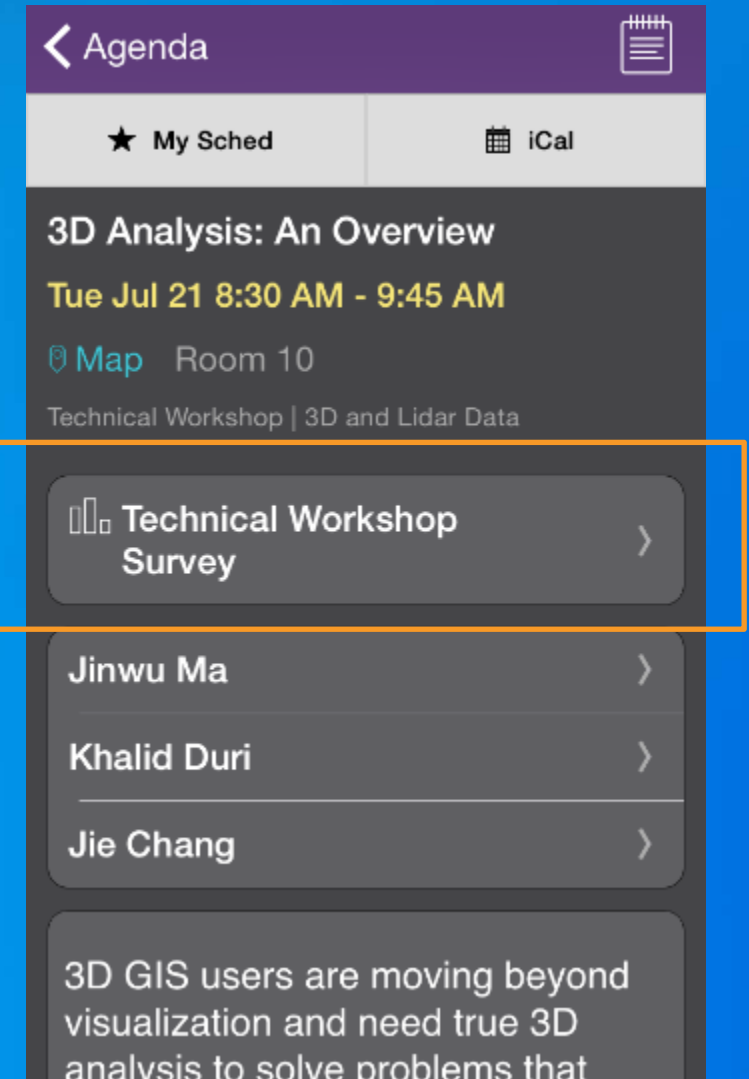
Recent papers

- Baella B, Lee D, Lleopart A, Pla M (2014) ICGC MRDB for topographic data: first steps in the implementation, The 17th ICA Generalization Workshop, 2014, Vienna, Austria. http://generalisation.icaci.org/images/files/workshop/workshop2014/genemr2014_submission_8.pdf
- Lee D (2015) Using Conflation for Keeping Data Harmonized and Up-to-date, to be presented at the ICA-ISPRS Workshop on Generalisation and Multiple Representation, 2015, Rio de Janeiro, Brazil
- Lee D, Yang W, Ahmed N (2014) Conflation in Geoprocessing Framework - Case Studies, GEOProcessing, 2014, Barcelona, Spain. <http://goo.gl/iOoSGV>
- Lee D, Yang W, Ahmed N (2015) Improving Cross-border Data Reliability Through Edgematching, to be presented at The 27th International Cartographic Conference, 2015, Rio de Janeiro, Brazil
- Yang W, Lee D, and Ahmed N, “Pattern Based Feature Matching for Geospatial Data Conflation”, GEOProcessing, 2014, Barcelona, Spain. <http://goo.gl/JKGJbo>

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- **Select Introduction to Geoprocessing Conflation tools and Workflows in the Mobile App - use Search Feature to find this title)**
- **Click “Technical Workshop Survey”**
- **Answer a few short questions and enter your comments**

*Thank you for attending! 😊
Any questions, comments ...?*





Understanding our world.