

Estimating Bilateral Migration

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IPFP

- A common problem with bilateral migration data is that it is unavailable or outdated.
 - Often collected in censuses
- In some cases there are other data sources available that provide information on the in- and out-migration totals
 - Population registers

A. Auxiliary Data

Origin	Destination				
	A	B	C	D	Sum
A		100	30	70	200
B	50		45	5	100
C	60	35		40	135
D	20	25	20		65
Sum	130	160	95	115	500

B. Primary Data: Marginal Totals

Origin	Destination				
	A	B	C	D	Sum
A					250
B					75
C					125
D					150
Sum	150	200	50	200	600

IPFP

- This provides a data estimation challenge, where the marginal tables totals are known but the cell values are known.
- Similar data estimation challenges exist for more detailed migration flow tables, for example:
 - In- and out-migration totals by age in each region are known, but the origin-destination migration flow table for each age group is missing.
 - Required by multi-regional cohort-component models
 - Estimating international migration flows from stocks (see for example Abel (2013))

IPFP

- A popular approach to estimate values in a contingency table based on known marginal tables and an initial contingency table is the Iterative Proportional Fitting Procedure (IPFP).
- First described by Deming and Stephan (1940), the IPFP has since been widely studied in a number of different disciplines and under a number of synonyms such as raking, matrix scaling or the RAS algorithm
 - Lovelace et al. (2015) gives a good overview of the application of IPFP in social sciences.
 - Lomax and Norman (2016) for another overview more specific to demography.
- Mathematical approach to iteratively adjust a *seed* contingency table $\mu_{ij}^{(0)} = m_{ij}$ to known row and column totals (n_{i+} and n_{+j})

$$\mu_{ij}^{(t+1)} = \frac{\mu_{ij}^{(t)}}{\mu_{i+}^{(t)}} n_{i+} \quad \mu_{ij}^{(t+2)} = \frac{\mu_{ij}^{(t+1)}}{\mu_{+j}^{(t+1)}} n_{+j}$$

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>				
	A	B	C	D	Sum
A		100	30	70	200
B	50		45	5	100
C	60	35		40	135
D	20	25	20		65
Sum	130	160	95	115	500

Totals

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>
	A	B	C	D	Sum	In
A		100	30	70	200	130
B	50		45	5	100	160
C	60	35		40	135	95
D	20	25	20		65	115
Sum	130	160	95	115	500	500

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>				<i>Sum</i>
	A	B	C	D	
A		100	30	70	200
B	50		45	5	100
C	60	35		40	135
D	20	25	20		65
Sum	130	160	95	115	500

<i>Totals</i>	
In	Out
130	200
160	100
95	135
115	65
500	500

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>				<i>Sum</i>
	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	
A		100	30	70	200
B	50		45	5	100
C	60	35		40	135
D	20	25	20		65
Sum	130	160	95	115	500

<i>Totals</i>		
<i>In</i>	<i>Out</i>	<i>Net</i>
130	200	-70
160	100	60
95	135	-40
115	65	50
500	500	0

IPFP

A. Auxiliary Data

Origin	Destination					Totals		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

[illegible]

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A						150	250			
B						200	75			
C						50	125			
D						200	150			
Sum						600	600			

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		100	30	70	200	150	250	130	200	-70
B	50		45	5	100	200	75	160	100	60
C	60	35		40	135	50	125	95	135	-40
D	20	25	20		65	200	150	115	65	50
Sum	130	160	95	115	500	600	600	500	500	0

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		125.00	37.50	87.50	250.00	150	250	139.21	250.00	-110.79
B	37.50		33.75	3.75	75.00	200	75	215.10	75.00	140.10
C	55.56	32.41		37.04	125.00	50	125	117.40	125.00	-7.60
D	46.15	57.69	46.15		150.00	200	150	128.29	150.00	-21.71
Sum	139.21	215.10	117.40	128.29	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		116.23	15.97	136.41	268.61	150	250	150.00	268.61	-118.61
B	40.41		14.37	5.85	60.63	200	75	200.00	60.63	139.37
C	59.86	30.13		57.74	147.74	50	125	50.00	147.74	-97.74
D	49.73	53.64	19.66		123.03	200	150	200.00	123.03	76.97
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		108.17	14.86	126.96	250.00	150	250	161.27	250.00	-88.73
B	49.99		17.78	7.23	75.00	200	75	199.07	75.00	124.07
C	50.65	25.50		48.86	125.00	50	125	56.61	125.00	-68.39
D	60.63	65.40	23.96		150.00	200	150	183.05	150.00	33.05
Sum	161.27	199.07	56.61	183.05	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		108.68	13.13	138.72	260.53	150	250	150.00	260.53	-110.53
B	46.49		15.70	7.90	70.10	200	75	200.00	70.10	129.90
C	47.11	25.61		53.38	126.10	50	125	50.00	126.10	-76.10
D	56.40	65.71	21.17		143.27	200	150	200.00	143.27	56.73
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		104.29	12.60	133.11	250.00	150	250	155.49	250.00	-94.51
B	49.74		16.80	8.45	75.00	200	75	198.47	75.00	123.47
C	46.70	25.39		52.91	125.00	50	125	51.56	125.00	-73.44
D	59.05	68.79	22.16		150.00	200	150	194.48	150.00	44.48
Sum	155.49	198.47	51.56	194.48	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		105.09	12.22	136.89	254.20	150	250	150.00	254.20	-104.20
B	47.99		16.29	8.69	72.98	200	75	200.00	72.98	127.02
C	45.05	25.59		54.41	125.05	50	125	50.00	125.05	-75.05
D	56.96	69.32	21.49		147.78	200	150	200.00	147.78	52.22
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		103.35	12.01	134.63	250.00	150	250	152.17	250.00	-97.83
B	49.32		16.75	8.94	75.00	200	75	199.30	75.00	124.30
C	45.03	25.58		54.39	125.00	50	125	50.57	125.00	-74.43
D	57.82	70.37	21.81		150.00	200	150	197.96	150.00	47.96
Sum	152.17	199.30	50.57	197.96	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		103.72	11.88	136.02	251.62	150	250	150.00	251.62	-101.62
B	48.62		16.56	9.03	74.20	200	75	200.00	74.20	125.80
C	44.39	25.67		54.95	125.01	50	125	50.00	125.01	-75.01
D	56.99	70.62	21.57		149.18	200	150	200.00	149.18	50.82
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		103.05	11.80	135.15	250.00	150	250	150.84	250.00	-99.16
B	49.14		16.73	9.13	75.00	200	75	199.72	75.00	124.72
C	44.39	25.66		54.95	125.00	50	125	50.22	125.00	-74.78
D	57.31	71.01	21.69		150.00	200	150	199.22	150.00	49.22
Sum	150.84	199.72	50.22	199.22	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		103.20	11.75	135.67	250.62	150	250	150.00	250.62	-100.62
B	48.87		16.66	9.16	74.69	200	75	200.00	74.69	125.31
C	44.14	25.70		55.16	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.10	21.59		149.69	200	150	200.00	149.69	50.31
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.94	11.72	135.34	250.00	150	250	150.32	250.00	-99.68
B	49.07		16.73	9.20	75.00	200	75	199.89	75.00	124.89
C	44.14	25.70		55.16	125.00	50	125	50.09	125.00	-74.91
D	57.11	71.25	21.63		150.00	200	150	199.70	150.00	49.70
Sum	150.32	199.89	50.09	199.70	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		103.00	11.70	135.54	250.24	150	250	150.00	250.24	-100.24
B	48.97		16.70	9.21	74.88	200	75	200.00	74.88	125.12
C	44.04	25.71		55.25	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.29	21.60		149.88	200	150	200.00	149.88	50.12
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.90	11.69	135.41	250.00	150	250	150.12	250.00	-99.88
B	49.04		16.73	9.23	75.00	200	75	199.96	75.00	124.96
C	44.04	25.71		55.24	125.00	50	125	50.03	125.00	-74.97
D	57.04	71.35	21.62		150.00	200	150	199.88	150.00	49.88
Sum	150.12	199.96	50.03	199.88	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.92	11.68	135.49	250.09	150	250	150.00	250.09	-100.09
B	49.00		16.72	9.23	74.95	200	75	200.00	74.95	125.05
C	44.01	25.72		55.28	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.36	21.60		149.95	200	150	200.00	149.95	50.05
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.88	11.68	135.44	250.00	150	250	150.05	250.00	-99.95
B	49.03		16.73	9.24	75.00	200	75	199.98	75.00	124.98
C	44.01	25.72		55.28	125.00	50	125	50.01	125.00	-74.99
D	57.01	71.39	21.61		150.00	200	150	199.96	150.00	49.96
Sum	150.05	199.98	50.01	199.96	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.89	11.67	135.47	250.04	150	250	150.00	250.04	-100.04
B	49.02		16.72	9.24	74.98	200	75	200.00	74.98	125.02
C	43.99	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.39	21.60		149.98	200	150	200.00	149.98	50.02
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.45	250.00	150	250	150.02	250.00	-99.98
B	49.03		16.73	9.24	75.00	200	75	199.99	75.00	124.99
C	43.99	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.40	21.60		150.00	200	150	199.98	150.00	49.98
Sum	150.02	199.99	50.00	199.98	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.88	11.67	135.46	250.01	150	250	150.00	250.01	-100.01
B	49.02		16.73	9.24	74.99	200	75	200.00	74.99	125.01
C	43.99	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.40	21.60		149.99	200	150	200.00	149.99	50.01
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.46	250.00	150	250	150.01	250.00	-99.99
B	49.03		16.73	9.24	75.00	200	75	200.00	75.00	125.00
C	43.99	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.41	21.60		150.00	200	150	199.99	150.00	49.99
Sum	150.01	200.00	50.00	199.99	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.46	250.01	150	250	150.00	250.01	-100.01
B	49.03		16.73	9.24	75.00	200	75	200.00	75.00	125.00
C	43.99	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.41	21.60		150.00	200	150	200.00	150.00	50.00
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.46	250.00	150	250	150.00	250.00	-100.00
B	49.03		16.73	9.24	75.00	200	75	200.00	75.00	125.00
C	43.99	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.41	21.60		150.00	200	150	200.00	150.00	50.00
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.46	250.00	150	250	150.00	250.00	-100.00
B	49.03		16.73	9.25	75.00	200	75	200.00	75.00	125.00
C	43.98	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.41	21.60		150.00	200	150	200.00	150.00	50.00
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.46	250.00	150	250	150.00	250.00	-100.00
B	49.03		16.73	9.25	75.00	200	75	200.00	75.00	125.00
C	43.98	25.72		55.29	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.41	21.60		150.00	200	150	200.00	150.00	50.00
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

A. Auxiliary Data

<i>Origin</i>	<i>Destination</i>					<i>Totals</i>		
	A	B	C	D	Sum	In	Out	Net
A		100	30	70	200	130	200	-70
B	50		45	5	100	160	100	60
C	60	35		40	135	95	135	-40
D	20	25	20		65	115	65	50
Sum	130	160	95	115	500	500	500	0

B. Constrained Estimates

<i>Origin</i>	<i>Destination</i>					<i>Target</i>		<i>Totals</i>		
	A	B	C	D	Sum	In	Out	In	Out	Net
A		102.87	11.67	135.46	250.00	150	250	150.00	250.00	-100.00
B	49.03		16.73	9.25	75.00	200	75	200.00	75.00	125.00
C	43.98	25.72		55.30	125.00	50	125	50.00	125.00	-75.00
D	56.99	71.41	21.60		150.00	200	150	200.00	150.00	50.00
Sum	150.00	200.00	50.00	200.00	600.00	600	600	600.00	600.00	0.00

IPFP

- Willekens (1999) calls the seed data an *auxiliary* table and notes that it should be information on a variables related to migration.
 - Typically past migration flow data
 - Distances or travel costs between the origin-destination pairs have been used where no past data exists
 - Limited testing to see which seeds work best for estimating migration
- The marginal data is then known as *primary* data.
 - Partial observations on the number of migrations

mipfp

- The *mipfp* package by Barthélemy and Suesse (2018) implements IPFP in R using the `Ipfp()` function
- Can be used for multi-dimensional marginal constraint problems.
- Three inputs
 - `seed` a matrix of auxiliary data to aid estimation
 - `target.list` a list of dimensions that are being targeted (see next point)
 - `target.data` a list of targets related to `target.list`
- R numbers dimension of arrays with
 - 1 row
 - 2 column
 - 3 table
 - ...
- The `target.list` might involve
 - a single target, e.g. column totals `target.list = list(2)`
 - multiple targets, e.g. row and column totals `target.list = list(1, 2)`
 - sums over cells rather than margins of array, e.g. cells summed over tables `target.list = list(c(1, 2))`

mipfp

```

> r <- LETTERS[1:4]
> m0 <- matrix(data = c(0, 100, 30, 70,
+                        50, 0, 45, 5,
+                        60, 35, 0, 40,
+                        20, 25, 20, 0),
+              nrow = 4, ncol = 4, byrow = TRUE,
+              dimnames = list(orig = r, dest = r))
> addmargins(m0)
      dest
orig    A    B    C    D Sum
  A      0 100 30   70 200
  B     50   0 45    5 100
  C     60 35  0   40 135
  D     20 25 20    0  65
Sum   130 160 95 115 500

```

mipfp

```
> orig_tot <- c(250, 75, 125, 150)
> dest_tot <- c(150, 200, 50, 200)
> names(orig_tot) <- names(dest_tot) <- r
>
> orig_tot
  A    B    C    D
250  75 125 150
> dest_tot
  A    B    C    D
150 200  50 200
>
> # check sums are equal
> sum(orig_tot)
[1] 600
> sum(dest_tot)
[1] 600
```

mipfp

```
> library(mipfp)
> Ipfp(seed = m0, target.list = list(1, 2),
+       target.data = list(orig_tot, dest_tot))
```

Call:

```
Ipfp(seed = m0, target.list = list(1, 2), target.data = list(orig_tot,
  dest_tot))
```

Method: ipfp - convergence: TRUE

Estimates:

	dest				
orig	A	B	C	D	
A	0.00000	102.87046	11.67024	135.459297	
B	49.02778	0.00000	16.72686	9.245364	
C	43.98433	25.72033	0.00000	55.295339	
D	56.98789	71.40921	21.60290	0.000000	

mipfp

```

> # save the result
> y0 <- Ipfp(seed = m0, target.list = list(1, 2),
+           target.data = list(orig_tot, dest_tot))
>
> # view with totals
> addmargins(y0$x.hat)

```

	dest				
orig	A	B	C	D	Sum
A	0.00000	102.87046	11.67024	135.459297	250
B	49.02778	0.00000	16.72686	9.245364	75
C	43.98433	25.72033	0.00000	55.295339	125
D	56.98789	71.40921	21.60290	0.000000	150
Sum	150.00000	200.00000	50.00000	200.000000	600

Three dimensions

<i>Auxiliary Data - Low Skill</i>						<i>Primary Data</i>					
<i>Origin</i>	<i>Destination</i>					<i>Origin</i>	<i>Destination</i>				
	A	B	C	D	Sum		A	B	C	D	Sum
A		80	10	55	145	A					250
B	30		20	0	50	B					75
C	50	15		10	75	C					125
D	5	20	10		35	D					150
Sum	85	115	40	65	305	Sum	150	200	50	200	600

<i>Auxiliary Data - High Skill</i>					
<i>Origin</i>	<i>Destination</i>				
	A	B	C	D	Sum
A		20	20	15	55
B	20		25	5	50
C	10	20		30	60
D	15	5	10		30
Sum	45	45	55	50	195

IPFP More Complicated Data Situations

- The IPFP can be used for more complex data situations with more than two dimensions.
- Key to using the `mipfp()` function is setting the inputs for `target.data`.

```
> library(tidyverse)
> d <- expand_grid(orig = r, dest = r, skill = c("Low", "High")) %>%
+   mutate(flow = c(0, 0, 80, 20, 10, 20, 55, 15, 30, 20, 0, 0, 20, 25, 0, 5, 50, 1
+
>
> d
# A tibble: 32 x 4
   orig dest skill flow
  <chr> <chr> <chr> <dbl>
1 A     A     Low     0
2 A     A     High    0
3 A     B     Low    80
4 A     B     High   20
5 A     C     Low   10
6 A     C     High   20
7 A     D     Low   55
8 A     D     High   15
9 B     A     Low   30
10 B    A     High   20
# ... with 22 more rows
```

Estimating Detailed Bilateral Migration

```
> m1 <- xtabs(formula = flow ~ orig + dest + skill, data = d)
```

```
> m1
```

```
, , skill = High
```

	dest			
orig	A	B	C	D
A	0	20	20	15
B	20	0	25	5
C	10	20	0	30
D	15	5	10	0

```
, , skill = Low
```

	dest			
orig	A	B	C	D
A	0	80	10	55
B	30	0	20	0
C	50	15	0	10
D	5	20	10	0

mipfp

```
> addmargins(m1)
```

```
, , skill = High
```

	dest				
orig	A	B	C	D	Sum
A	0	20	20	15	55
B	20	0	25	5	50
C	10	20	0	30	60
D	15	5	10	0	30
Sum	45	45	55	50	195

```
, , skill = Low
```

	dest				
orig	A	B	C	D	Sum
A	0	80	10	55	145
B	30	0	20	0	50
C	50	15	0	10	75
D	5	20	10	0	35
Sum	85	115	40	65	305

```
, , skill = Sum
```

	dest				
orig	A	B	C	D	Sum

mipfp

```
> addmargins(m1)[, , skill = "Sum"]
```

```
dest
```

orig	A	B	C	D	Sum
A	0	100	30	70	200
B	50	0	45	5	100
C	60	35	0	40	135
D	20	25	20	0	65
Sum	130	160	95	115	500

mipfp

```
> y1 <- Ipfp(seed = m1, target.list = list(1, 2),
+           target.data = list(orig_tot, dest_tot))
> addmargins(y1$x.hat)
, , skill = High
```

	dest				
orig	A	B	C	D	Sum
A	0.000000	20.574092	7.780161	29.026992	57.381245
B	19.611112	0.000000	9.292698	9.245364	38.149174
C	7.330722	14.697330	0.000000	41.471504	63.499556
D	42.740914	14.281842	10.801451	0.000000	67.824208
Sum	69.682749	49.553264	27.874310	79.743860	226.854183

```
, , skill = Low
```

	dest				
orig	A	B	C	D	Sum
A	0.000000	82.296369	3.890080	106.432305	192.618755
B	29.416668	0.000000	7.434158	0.000000	36.850826
C	36.653611	11.022997	0.000000	13.823835	61.500444
D	14.246971	57.127369	10.801451	0.000000	82.175792
Sum	80.317251	150.446736	22.125690	120.256140	373.145817

```
, , skill = Sum
```

mipfp

```
> addmargins(y1$x.hat)[, , skill = "Sum"]
```

```
dest
```

orig	A	B	C	D	Sum
A	0.000000	102.870462	11.670241	135.459297	250.000000
B	49.027781	0.000000	16.726856	9.245364	75.000000
C	43.984334	25.720327	0.000000	55.295339	125.000000
D	56.987886	71.409211	21.602903	0.000000	150.000000
Sum	150.000000	200.000000	50.000000	200.000000	600.000000

mipfp

```
> y1$x.hat %>%  
+   as.data.frame.table(responseName = "est") %>%  
+   as_tibble()  
# A tibble: 32 x 4  
  orig dest skill    est  
  <fct> <fct> <fct> <dbl>  
1 A     A     High    0  
2 B     A     High  19.6  
3 C     A     High   7.33  
4 D     A     High  42.7  
5 A     B     High  20.6  
6 B     B     High    0  
7 C     B     High  14.7  
8 D     B     High  14.3  
9 A     C     High   7.78  
10 B    C     High   9.29  
# ... with 22 more rows
```

Net constrained origin-destination flows

- Plane (1981) developed a proportional adjustment algorithm for estimating bilateral migration flows to match both
 - Constraints on the net migration of each region
 - Total sum of the bilateral migration flows
- Requires knowledge of
 - Past bilateral migration flows
 - Current (target) total migration flows (over whole system)
 - Current (target) net migration flows
 - Distance matrix to correspond
- No application of this method in R, although in *migest* package the `cm_net_tot()` function provides a similar set of estimates
 - Unable to incorporate distance matrix

Net constrained origin-destination flows

```
> addmargins(m0)
      dest
orig    A    B    C    D Sum
  A      0 100 30   70 200
  B     50   0 45    5 100
  C     60 35  0   40 135
  D     20 25 20    0  65
Sum    130 160 95 115 500
```

```
>
> # observed net
> library(migest)
> sum_turnover(m0)
# A tibble: 4 x 5
  region in_mig out_mig turn    net
  <chr>   <dbl>   <dbl> <dbl> <dbl>
1 A           130       200   330   -70
2 B           160       100   260    60
3 C            95       135   230  -40
4 D           115        65   180   50
```

Net constrained origin-destination flows

- Estimate migration flows to match new net migration and grand total.

```
> y1 <- cm_net_tot(net_tot = c(-100, 125, -75, 50), tot = 600,
+                  m = m0, verbose = FALSE)
> addmargins(y1$n)
```

```
      dest
orig    A      B      C      D      Sum
A      0.00000 136.22513 32.93756 79.068944 248.23163
B      49.88761  0.00000 42.28296  4.833488  97.00406
C      74.27815 50.62851  0.00000 47.977516 172.88418
D      24.06590 35.15032 22.66377  0.000000  81.87999
Sum 148.23166 222.00396 97.88429 131.879947 599.99986
```

```
> sum_turnover(y1$n)
```

```
# A tibble: 4 x 5
```

```
  region in_mig out_mig turn  net
<chr>   <dbl>   <dbl> <dbl> <dbl>
1 A      148.    248.   396. -100.
2 B      222.    97.0   319.  125.
3 C       97.9   173.   271. -75.0
4 D      132.    81.9   214.  50.0
```

Net constrained origin-destination flows

- The requirement on the total sum of the bilateral flow for the algorithm is not realistic. (1981) method not widely adopted
 - In many countries the overall number of migrant flows, that is demographically consistent with natural population change, is typically not known.
 - If the overall number of migrant flows is known, it is typically obtained from a comprehensive population register, and thus bilateral migration or total in- and out-migration flows already exist. If it is the later, can use IPFP approaches.
- In recent years I have been working on a method that constrains only to the net migration totals.
 - Unpublished, work in progress, use at own risk
 - Method is available in the `cm_net()` function in the *migest* package
- Potential uses
 - Update bilateral migration flows from surveys or administrative data to match known demographic consistent net migration totals
 - Estimate bilateral migration flows from known net migration totals using non-migration data as a seed (if no migration flow data available)

Net constrained origin-destination flows

```
> y2 <- cm_net(net_tot = c(-100, 125, -75, 50), m = m0, verbose = FALSE)
```

```
> addmargins(y2$n)
```

```
      dest
```

orig	A	B	C	D	Sum
A	0.00000	124.97056	27.96585	71.121910	224.05832
B	40.00942	0.00000	33.56693	4.065067	77.64142
C	64.36422	46.92119	0.00000	43.597199	154.88260
D	19.68451	30.74980	18.34980	0.000000	68.78412
Sum	124.05815	202.64155	79.88258	118.784175	525.36645

```
> sum_turnover(y2$n)
```

```
# A tibble: 4 x 5
```

	region	in_mig	out_mig	turn	net
	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
1	A	124.	224.	348.	-100.
2	B	203.	77.6	280.	125.
3	C	79.9	155.	235.	-75.0
4	D	119.	68.8	188.	50.0

Exercise (ex7.R)

```
# 0.  a) Load the KOSTAT2021.Rproj file.
#      Run the getwd() below. It should print the directory where the
#      KOSTAT2021.Rproj file is located.
getwd()
#      b) Load the packages used in this exercise
library(tidyverse)
library(mipfp)
##
##
##
# 1. Run the code below to read in the bilateral data in uk_census_2011_tidy.csv
#      from the ONS 2011 British Census
cen11 <- read_csv("./data/uk_census_2011_tidy.csv")
cen11
# 2. Run the code below to read in the bilateral data in
#      uk_nhs_hesa_2018.csv from the British administrative data (National
#      Health Service patient records and Higher Education Statistics Authority)
nhs18 <- read_csv("./data/uk_nhs_hesa_2018_tidy.csv")
nhs18
# 3. Run the code below to create data with abbreviated region names - to make
#      it easier to view the matrices for each time period
#      Note: the census data is more detailed (orig - dest - age - sex) than the
#      administrative data (orig - dest)
cen11 <- cen11 %>%
  mutate(orig_full = orig,
```

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