

Using a Two-Part Markov Latent Class Model to Examine Expenditure Report Quality

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Latent Class Analysis

- Uses repeated measurements from panel survey data to estimate classification error
- Does not require external validation data; estimates of error directly from panel data
- LCA used to study measurement or response error (VandePol and deLeeuw 1986; Tucker 1992; Van de Pol and Langeheine 1997; Bassi et al. 2000; Biemer and Bushery 2000; Tucker, et al. 2002, 2003, 2004, 2005, 2006, and 2008); Meekins et al. (2011)

U.S. Consumer Expenditure Interview Survey (CEIS)

- ~ 6,000 CU's/year
- CU's interviewed every 3 months about prior 3 months expenditures
- 4 consecutive interviews on each CU
- 15 years of CEIS: 1996-2010
- Unweighted analysis
- 31 commodity categories analyzed

Commodity Categories

Dental

Computer games

Childcare

Prescription drugs

Computer equipment

Pets and pet supplies

Eye care

Books

Major Vehicle Repairs

Clothing

Cable

Minor Vehicle Repairs

Infant clothing

Music

License/registration

Clothing accessories

Internet (2001+)

HH electricity

Clothing services

Sports equipment

HH gas

Sewing

Major appliances

HH trash service

Shoes

Minor appliances

Phone

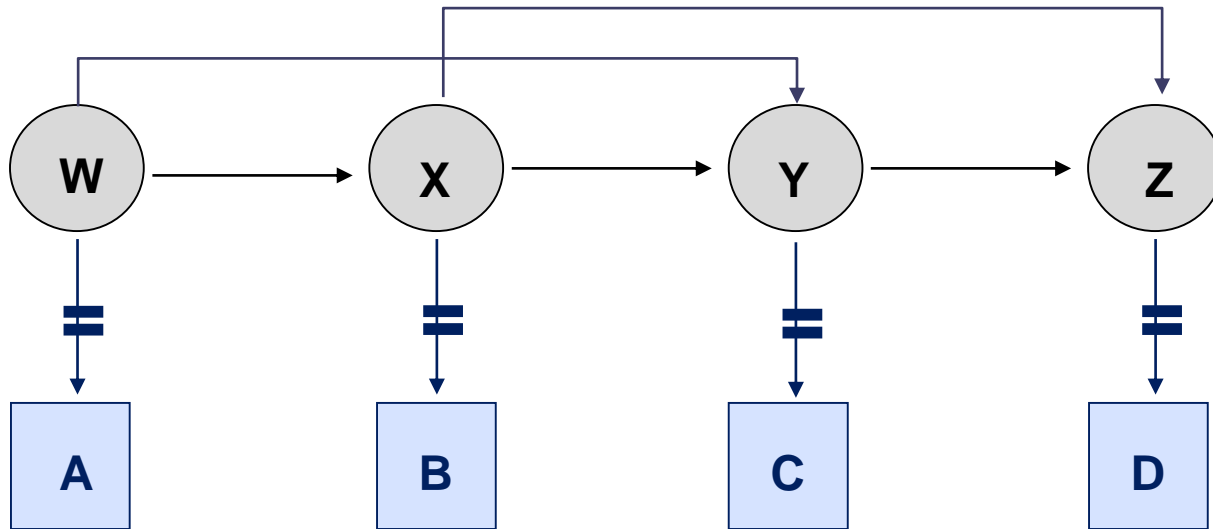
Jewelry

Electronics

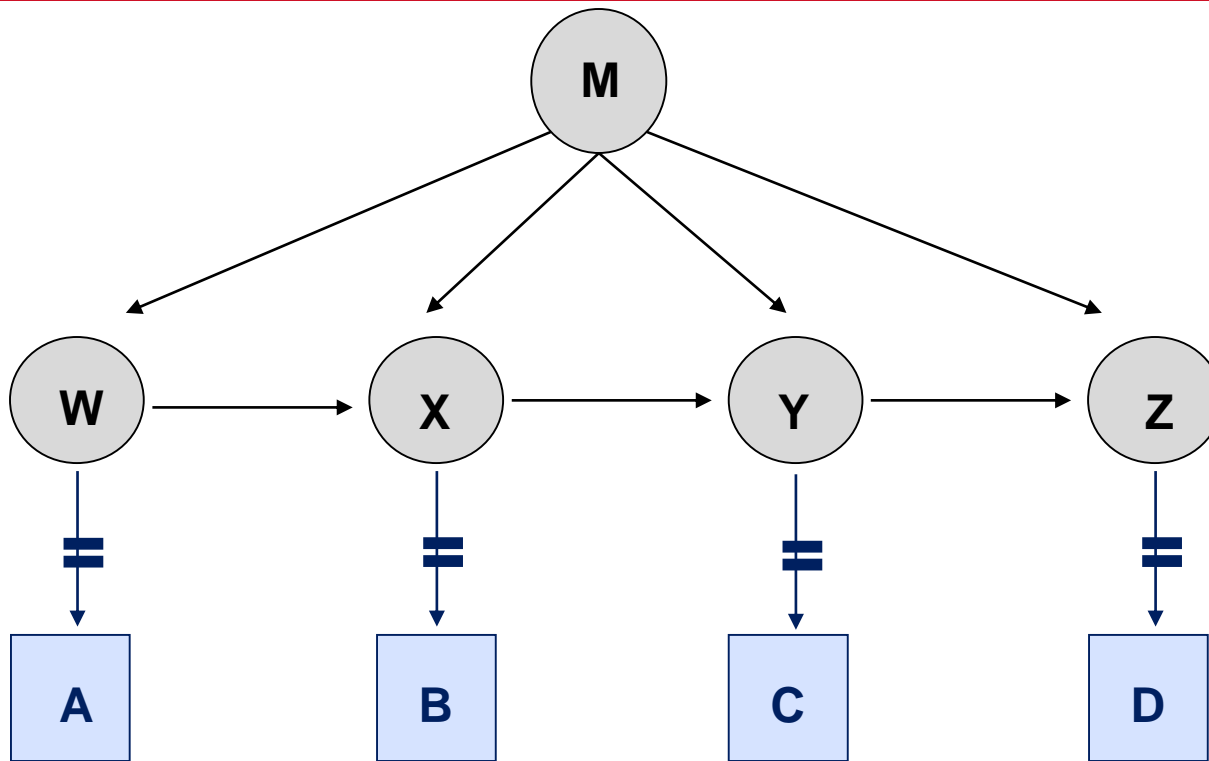
HH services

Events (e.g. sporting/theatre)

2nd Order Markov



Mover-Stayer



$$M = \begin{cases} 1, & P(W), P(X), P(Y), \text{ and } P(Z) \text{ are unconstrained.} \\ 2, & P(W=1) = P(X=1) = P(Y=1) = P(Z=1) = 1 \\ 3, & P(W=1) = P(X=1) = P(Y=1) = P(Z=1) = 0 \end{cases} \quad 6$$

Model Assumptions

- Markov or Mover-Stayer model assumptions
- Equal measurement error across all interviews

$$\begin{aligned} P(a_i = j | w_i = k) &= P(b_i = j | x_i = k) \\ &= P(c_i = j | y_i = k) = P(d_i = j | z_i = K) = q_{jk} \end{aligned}$$

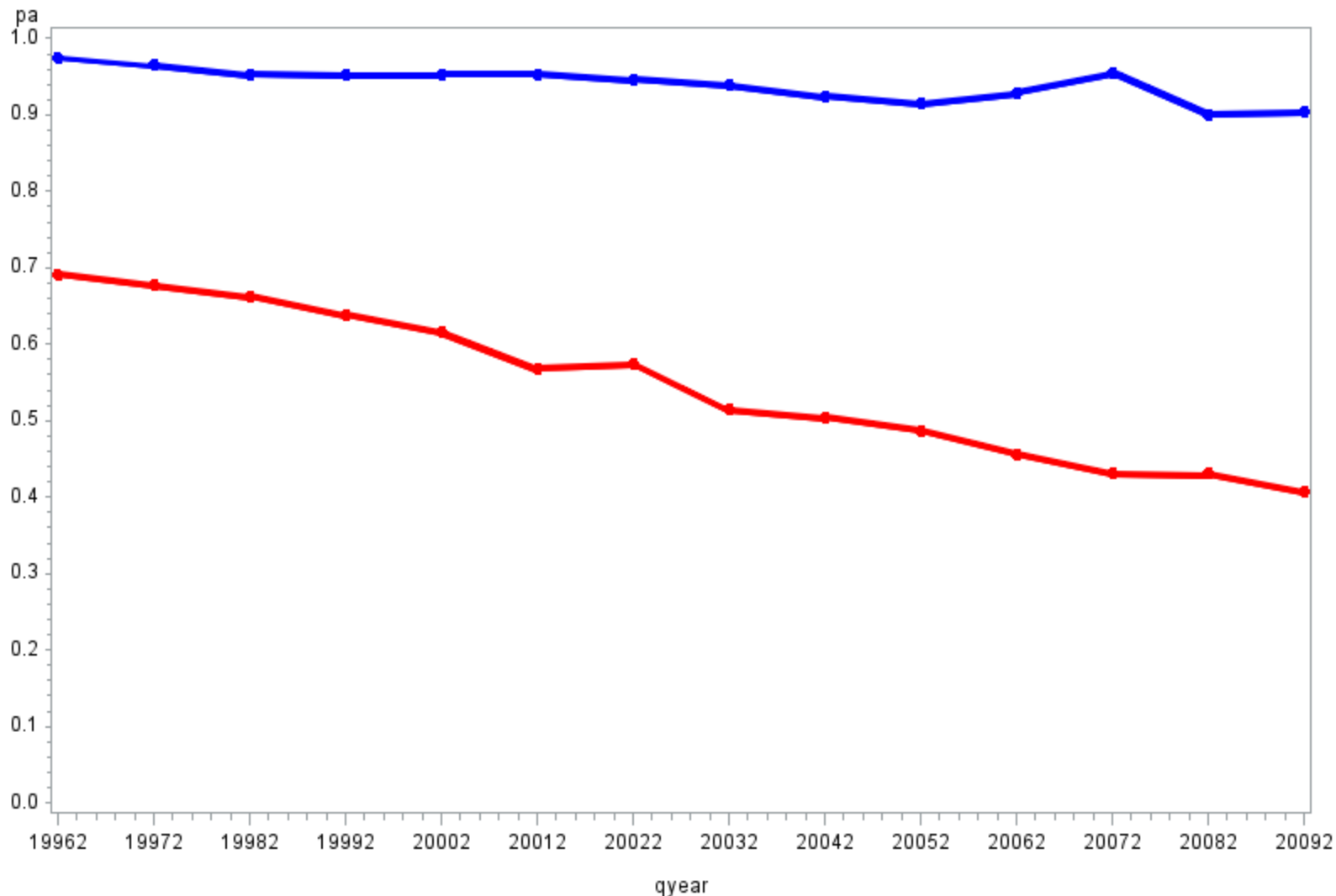
- No False Positives

$$P(a_i = 1 | w_i = 2) = 0$$

$$P(a_i = 2 | w_i = 2) = 1$$

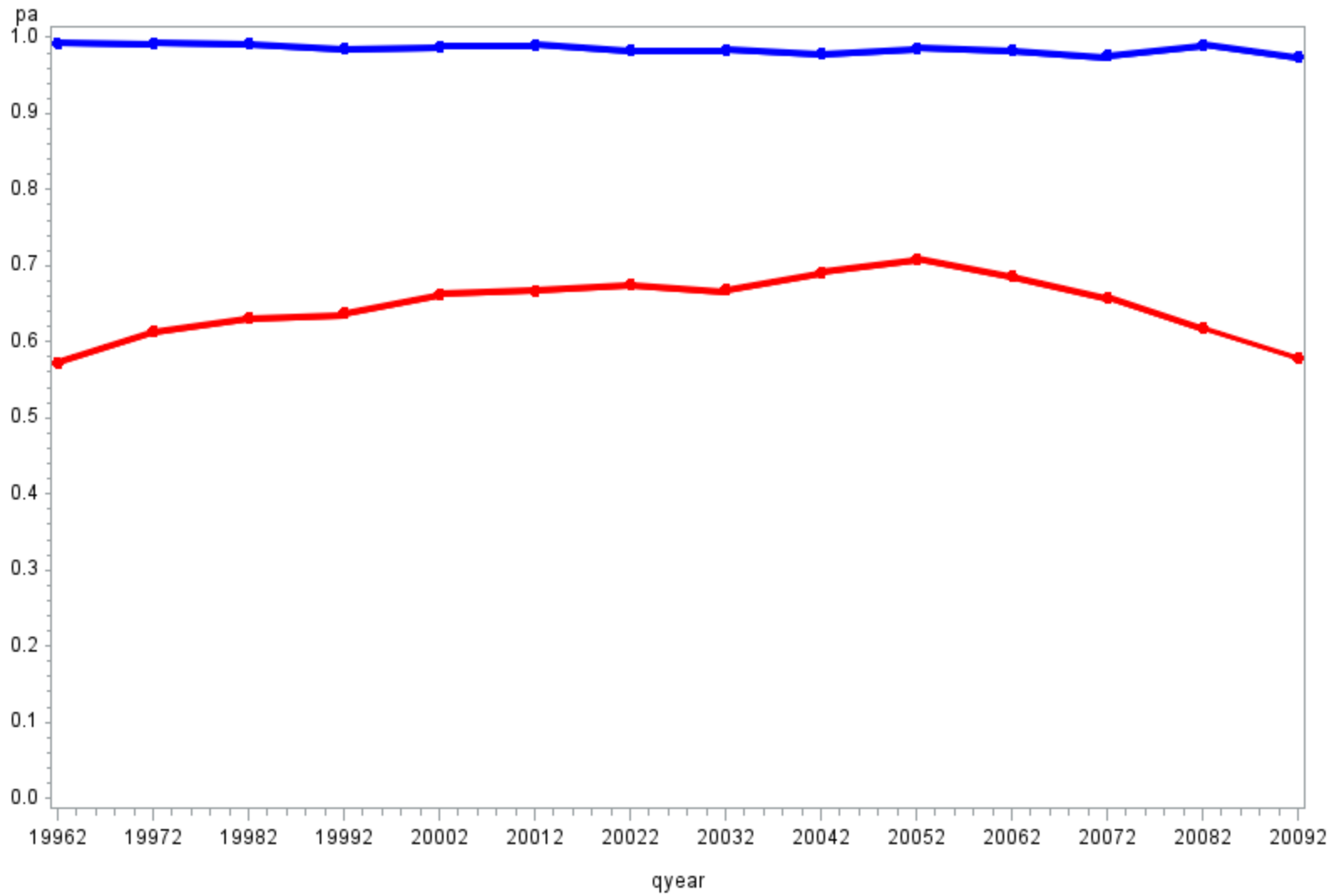
LCA MOVER-STAYER OVER TIME (1 YEAR POOLED COHORTS)

books by Year



PLOT ●—● Prob(A=1) ●—● Prob(A=1|W=1)

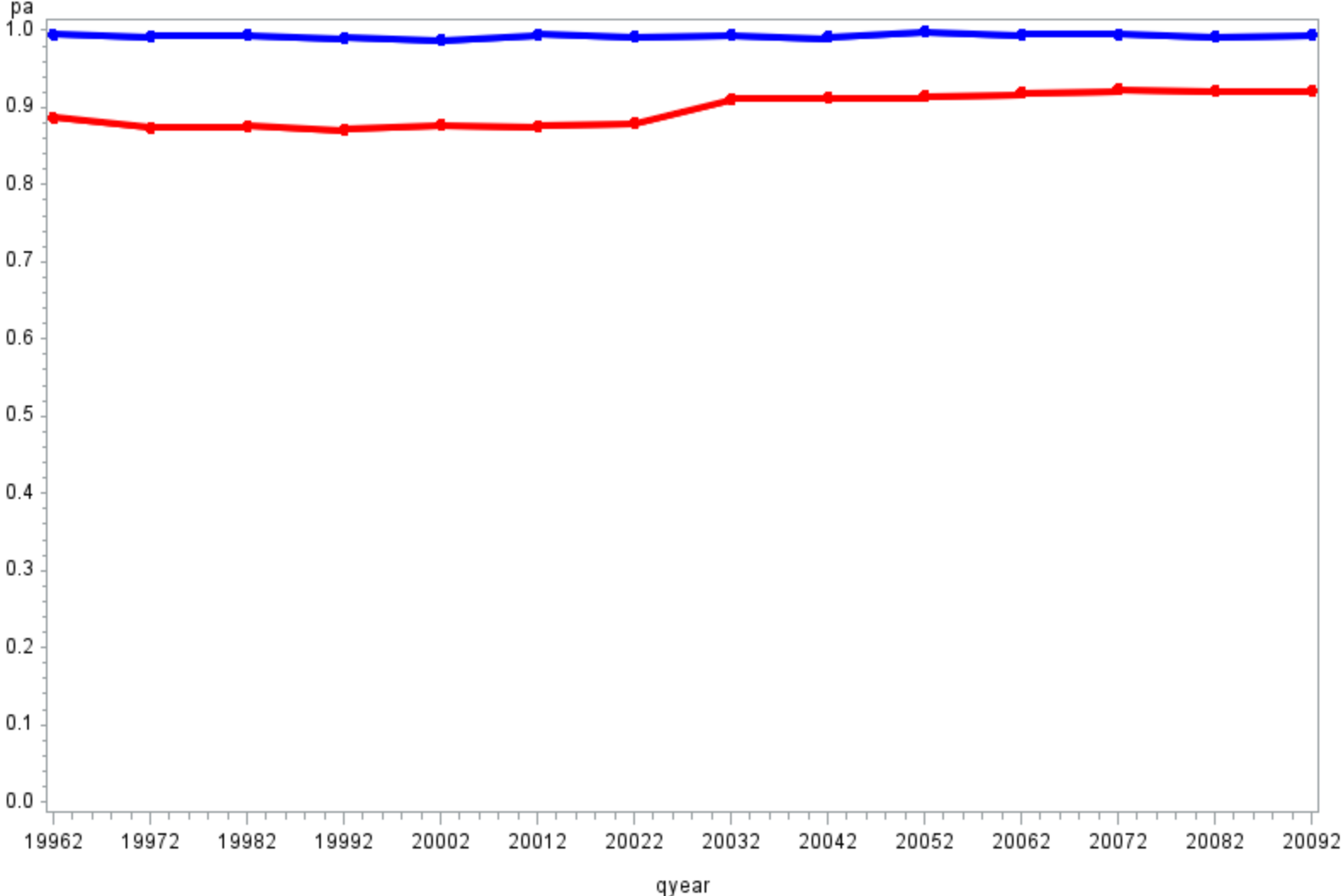
cable by Year



PLOT ●-●-● Prob(A=1) ●-●-● Prob(A=1|W=1)



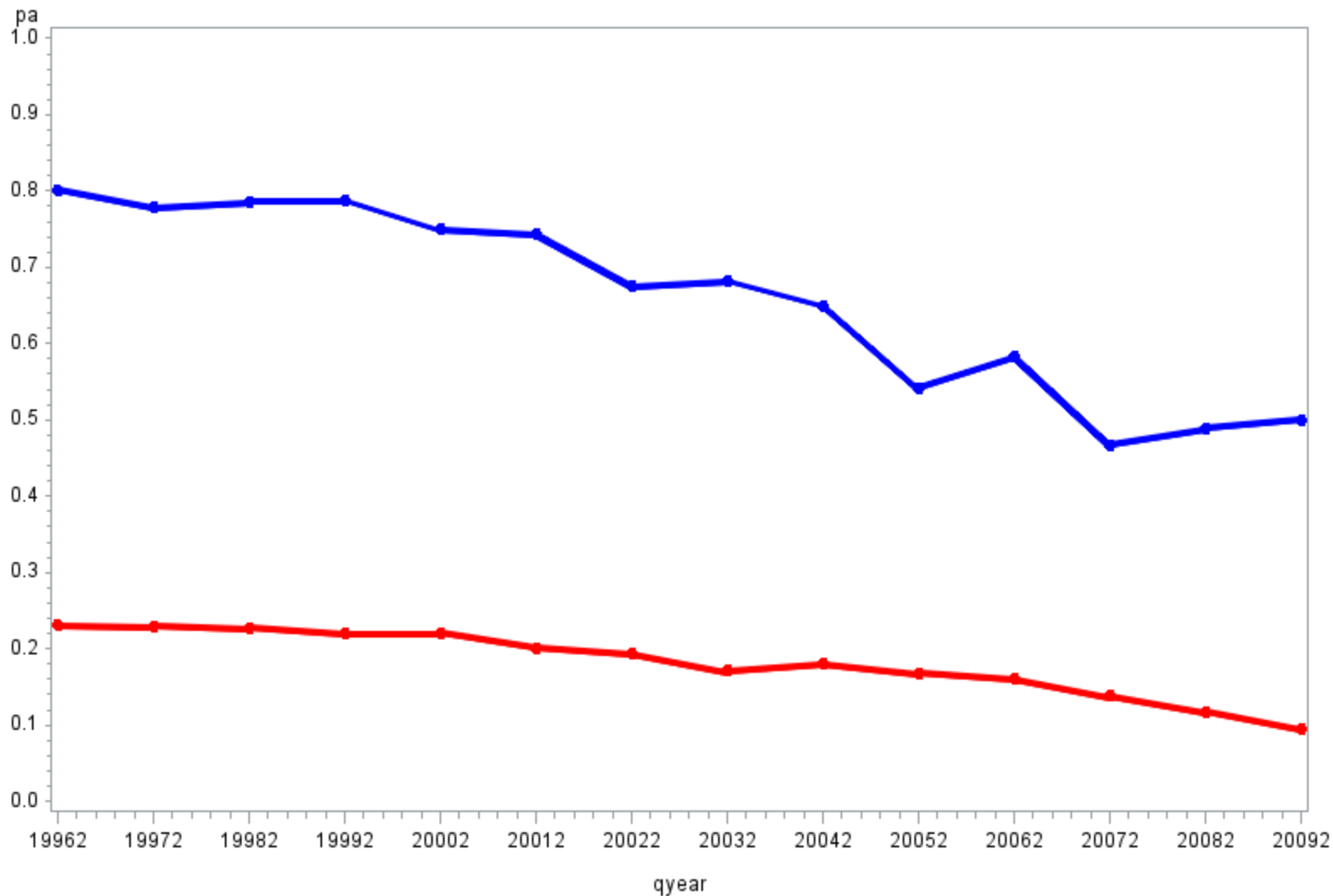
electric by Year



PLOT  Prob(A=1)  Prob(A=1|W=1)



music by Year



PLOT  Prob(A=1)  Prob(A=1|W=1)

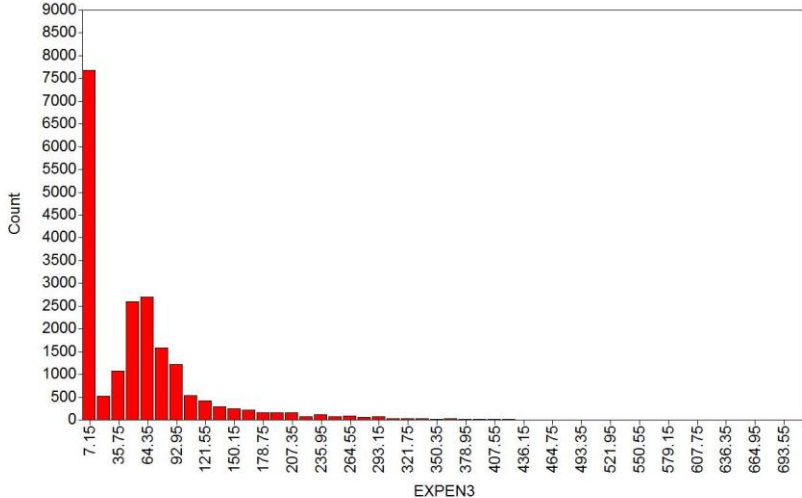
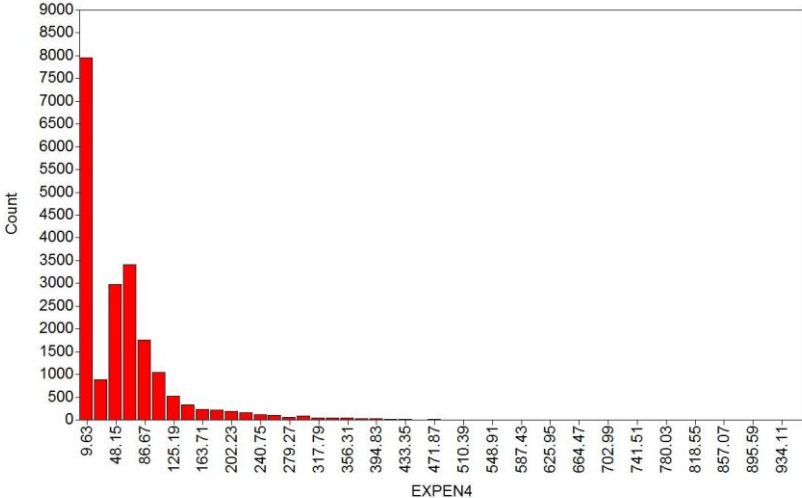
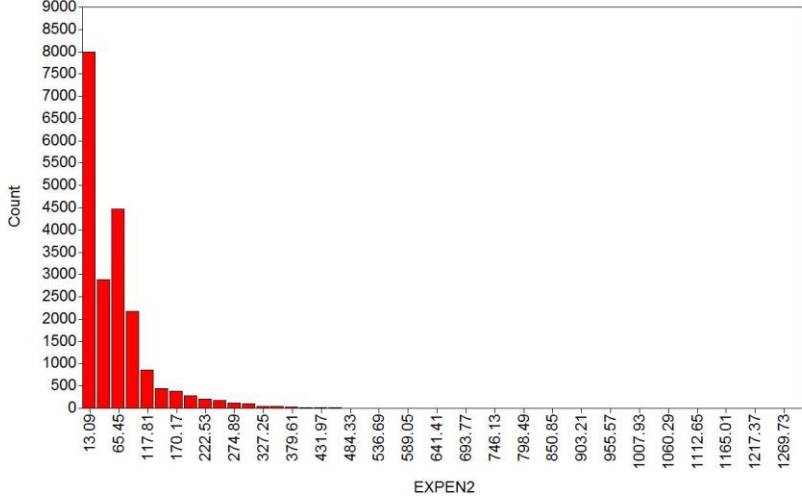
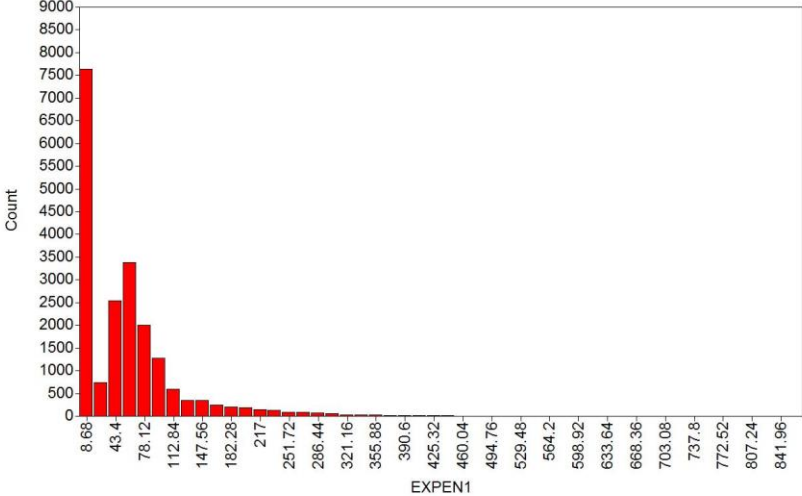
Summary of Previous Findings

- Assessment of MLCA for the detection of *change* in measurement error/time
- Accuracy rates/all estimates noisy
- Estimates are reasonable
- Useful: Sensitive to survey changes

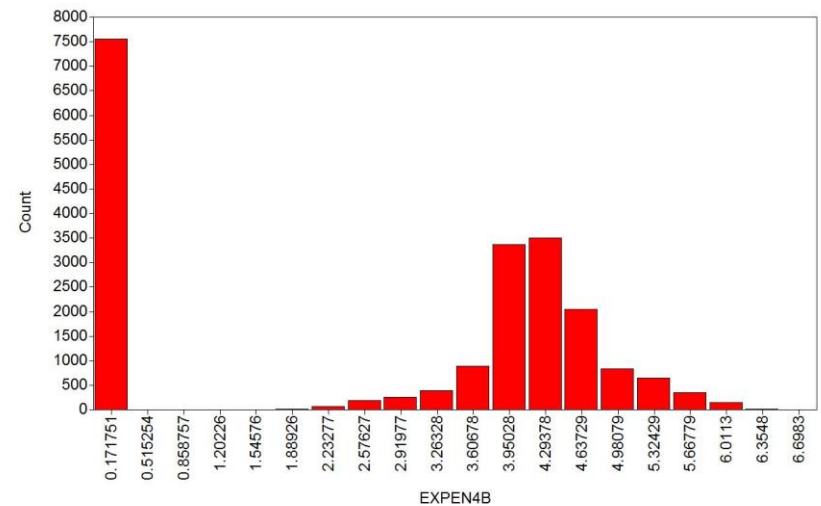
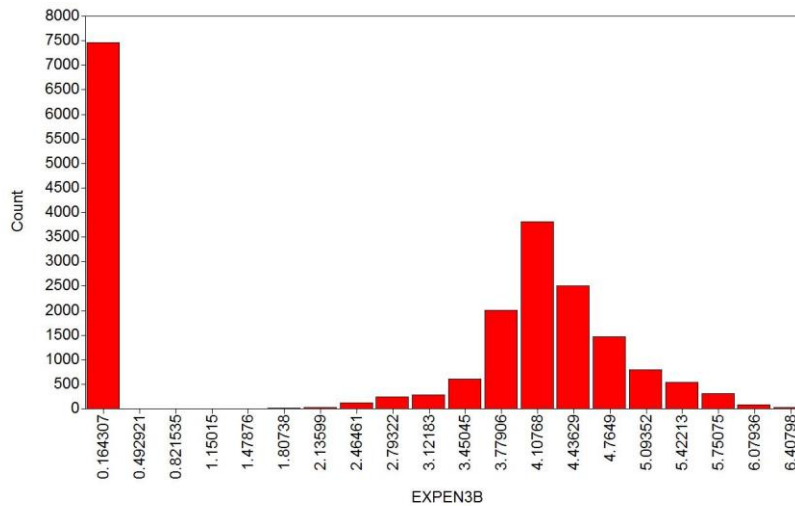
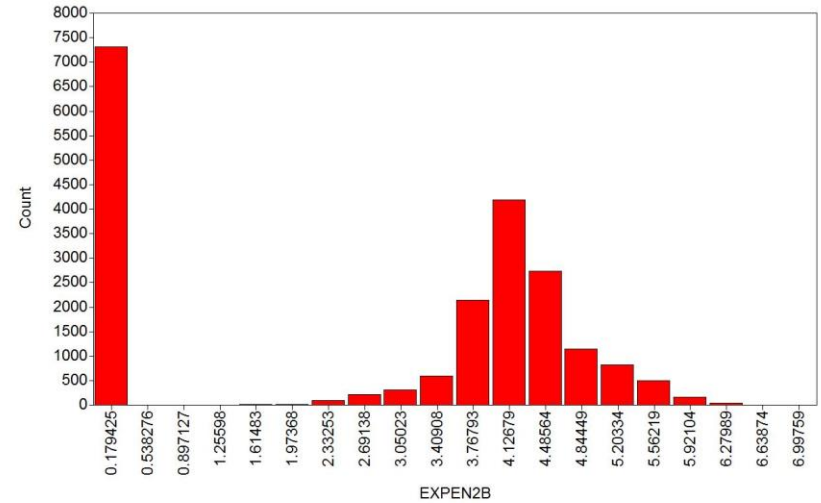
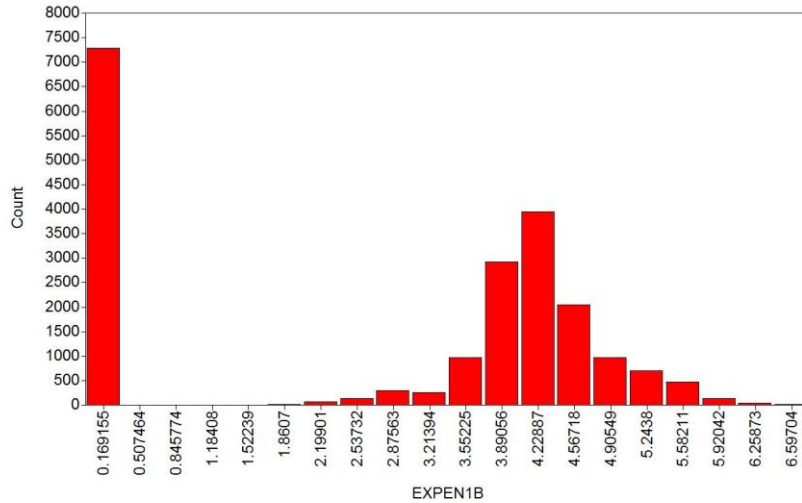
Add Expenditure

- Adds information to the model
- Allows for simultaneous (as opposed to two-stage estimation of unreported expenditure)
- If auto-correlative effects are large – stabilization of estimates should result

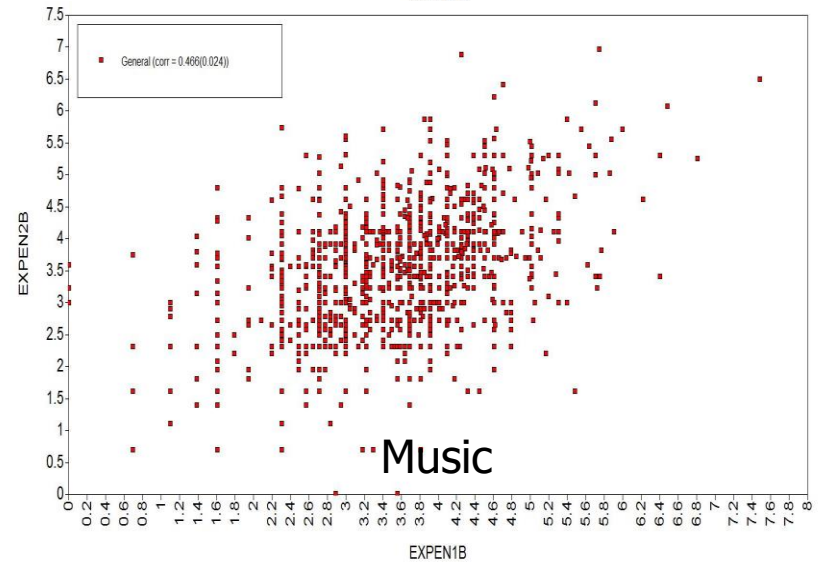
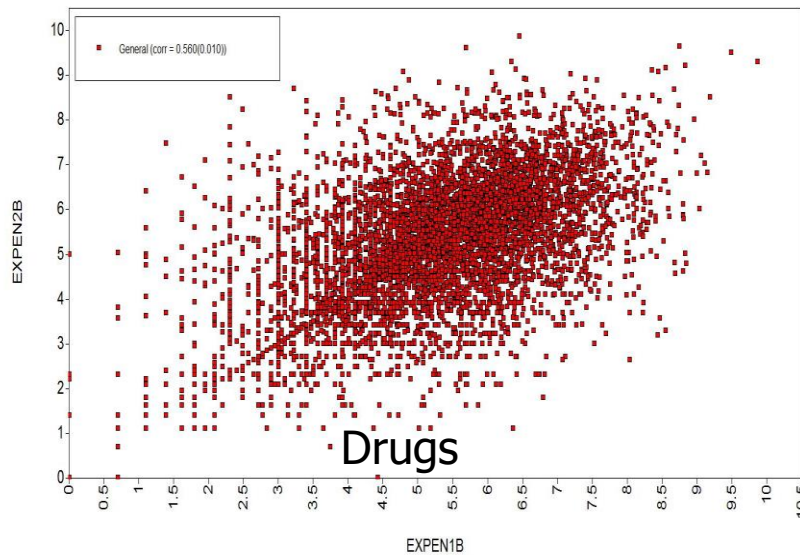
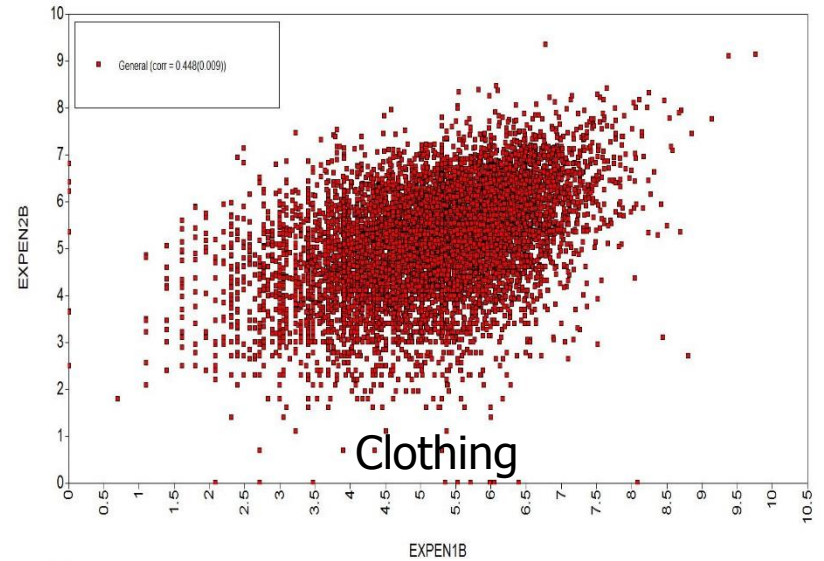
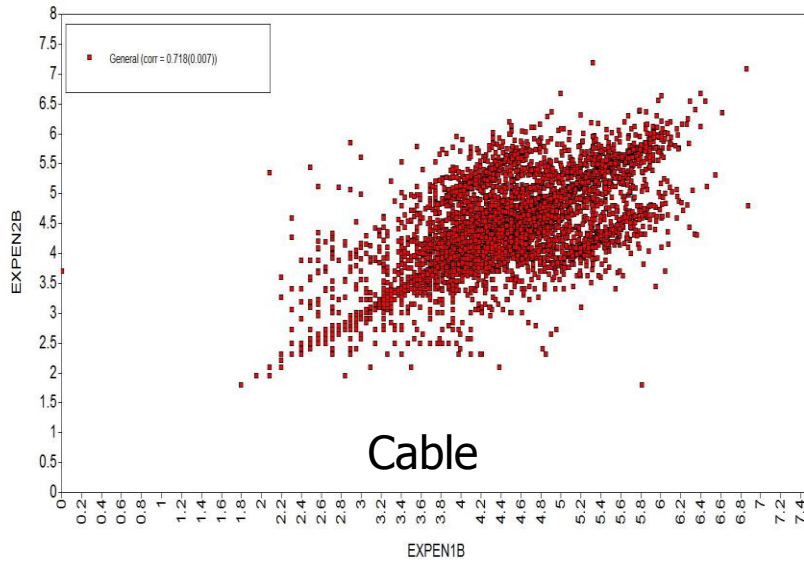
Expenditure data raw



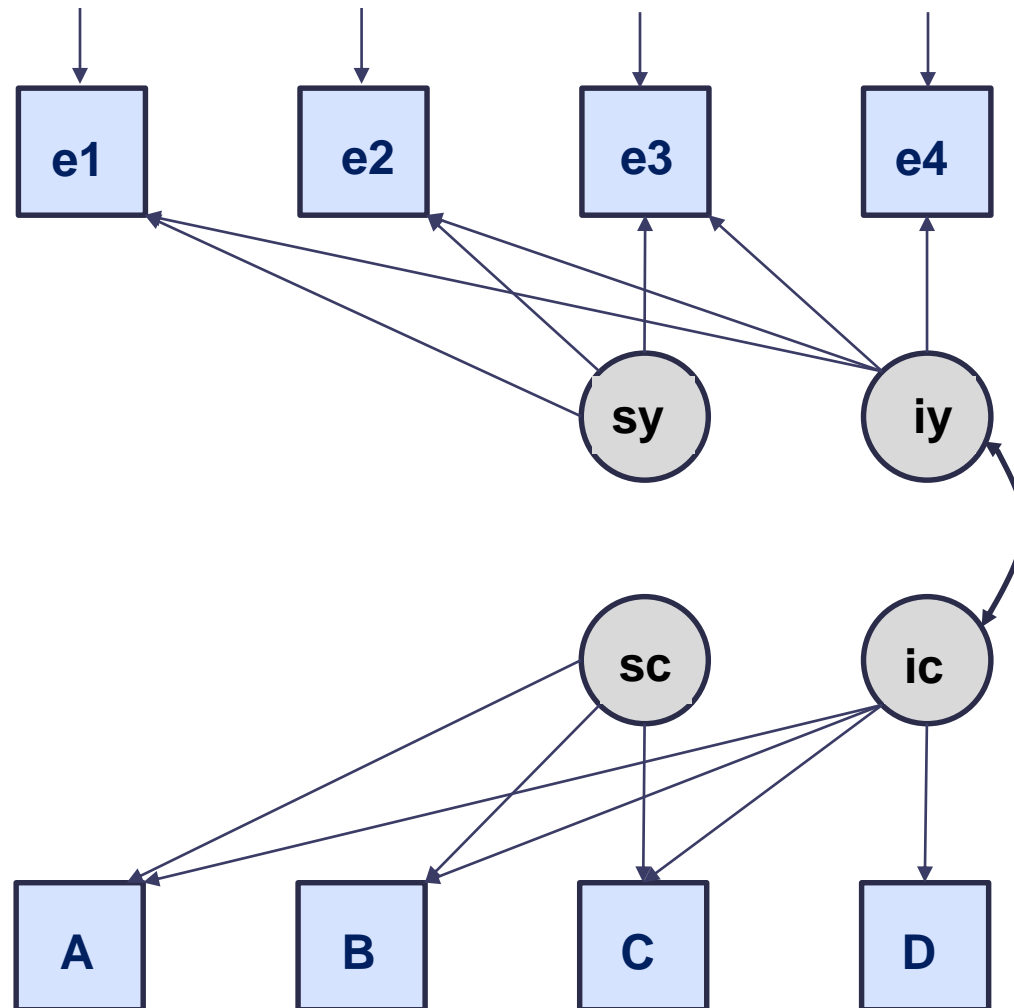
Expenditure Data (In)



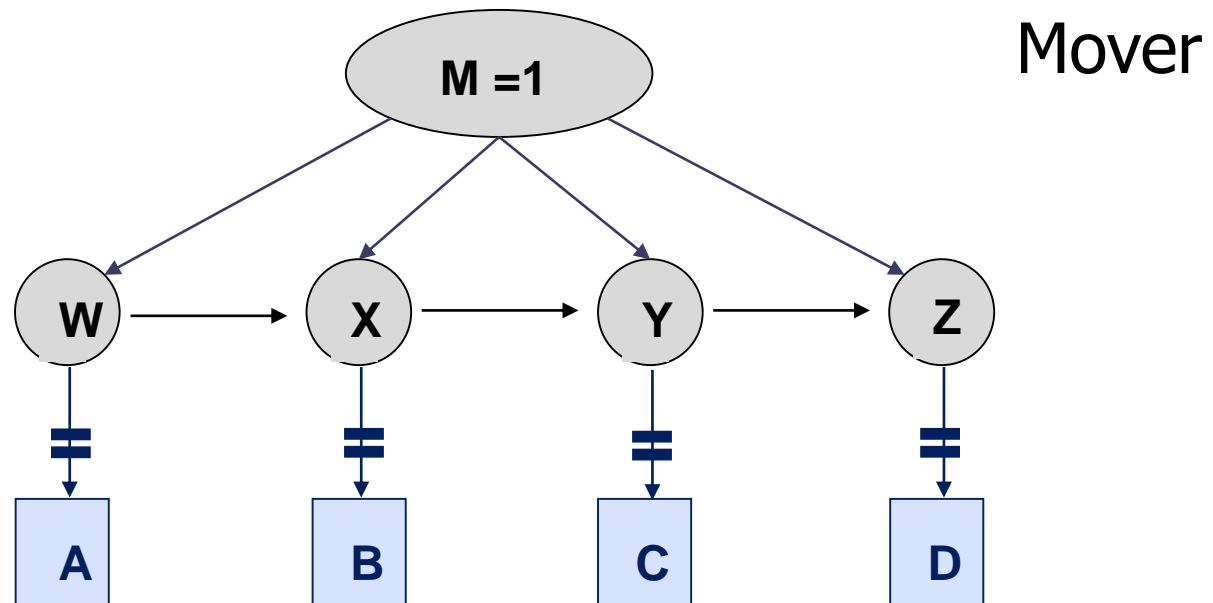
Autoregressive (Observed)



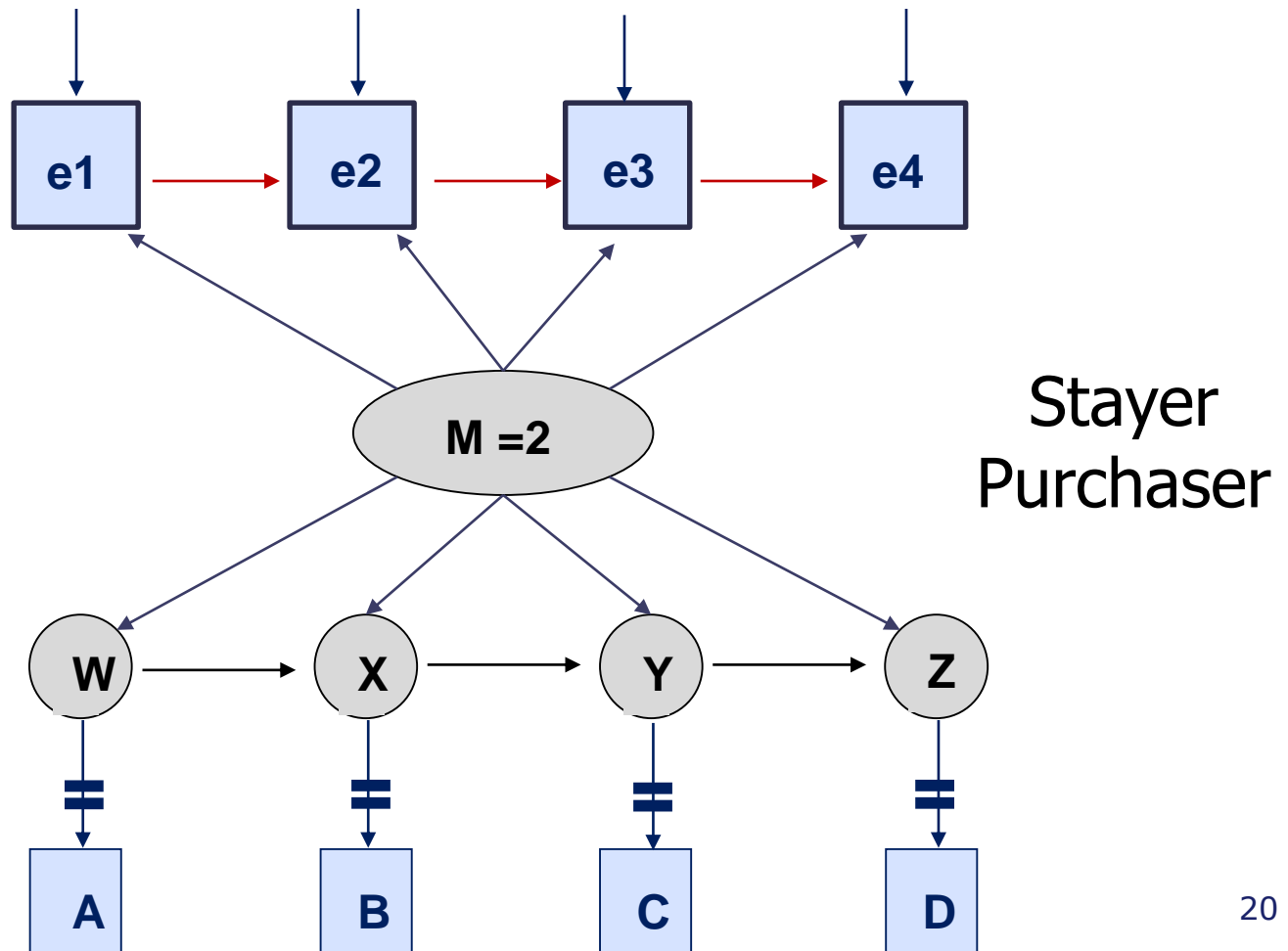
Two-part Model (Olsen and Schaeffer 2001)



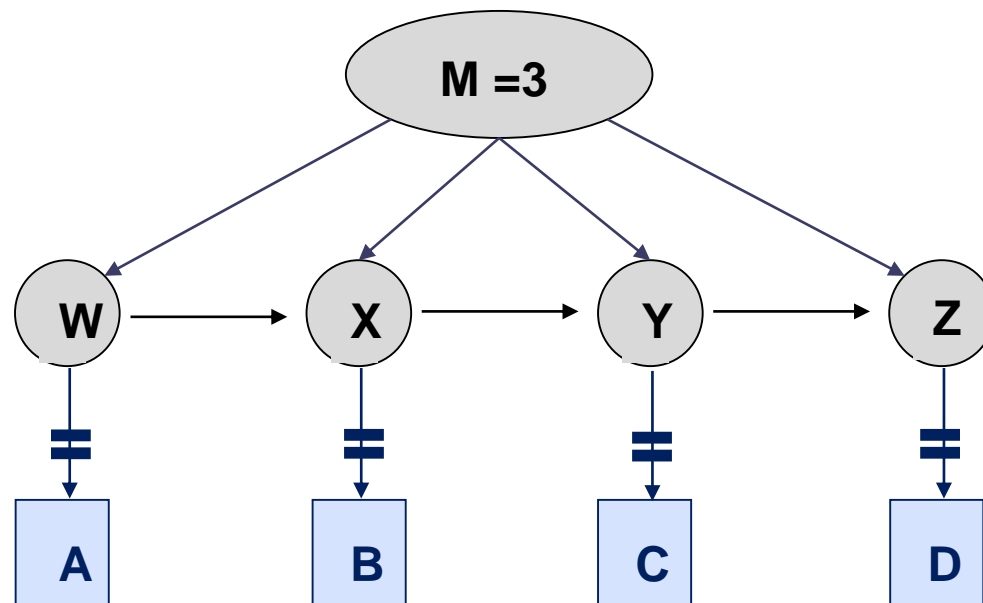
Modified MS



Modified MS



Modified MS



Stayer
Non-
Purchaser

Objective Diagnostics

■ Fit Statistics

▶ L-square
$$L^2 = 2 \sum_i n_i \ln \left(\frac{n_i}{\hat{m}_i} \right)$$

▶ Entropy
$$en(\alpha) = - \sum_i^N \sum_j^J \alpha_{ij} \log \alpha_{ij}$$

▶ BIC
$$BIC = L^2 - df \log N$$

Results Model Fit

Cable

Model	log L	BIC	Entropy
2 nd order	-34769.864	69579.396	.961
Mover-Stayer	-30532.432	61183.868	.845
Two part	-63092.649	126294.386	na
Modified MS	-148951.831	297949.663	.983

Results Model Fit

Clothing

Model	log L	BIC	Entropy
2 nd order	-50629.582	101298.83	.801
Mover-Stayer	-49623.976	99366.956	.793
Two part	-121605.028	243319.143	na
Modified MS	-227619.635	455285.270	.935

Results Model Fit

Drugs

Model	log L	BIC	Entropy
2 nd order	-48549.757	97139.181	.768
Mover-Stayer	-45002.673	90124.351	.766
Two part	-106225.730	212560.548	na
Modified MS	-214846.321	429920.735	.924

Results Model Fit

Major Appliances

Model	log L	BIC	Entropy
2 nd order	-16041.502	32122.673	.200
Mover-Stayer	-15990.318	32099.640	.576
Two part	-22532.573	45174.233	na
Modified MS	-123067.662	246363.416	.759

Results Model Fit

Music

Model	log L	BIC	Entropy
2 nd order	-32777.309	65594.285	.503
Mover-Stayer	-31134.973	62388.951	.769
Two part	-46334.975	92779.037	na
Modified MS	-135818.829	271865.751	.823

Results, Reports / Expenditure

Cable

Model	$P(A=1 / W=1)$	Missing Expenditure /QTR	Missing Expenditure /CU QTR
2 nd order	.984	\$1,751.96*	\$1.17*
Mover-Stayer	.984	\$1,751.96*	\$1.17*
Two part	na	\$18,358.56	\$12.24
Mod MS	.984	\$996.16	\$.66
<i>*Estimated from $P(A/W)$</i>			

Results, Reports / Expenditure

Clothing

Model	$P(A=1 / W=1)$	Missing Expenditure /QTR	Missing Expenditure /CU QTR
2 nd order	.910	\$25,059.97*	\$16.71*
Mover-Stayer	.846	\$46,124.29*	\$30.75*
Two part	na	-\$26,799.50	-\$17.87
Mod MS	.742	\$33,992.79	\$22.66
<i>*Estimated from $P(A/W)$</i>			

Results, Reports / Expenditure

Drugs

Model	$P(A=1 / W=1)$	Missing Expenditure /QTR	Missing Expenditure /CU QTR
2 nd order	.817	\$60,399.74*	\$40.27*
Mover-Stayer	.903	\$28,966.10*	\$19.31*
Two part	na	-\$33,316.50	-\$22.21
Mod MS	.877	\$17,222.49	\$11.48
<i>*Estimated from $P(A/W)$</i>			

Results, Reports / Expenditure

Major Appliances

Model	$P(A=1 / W=1)$	Missing Expenditure /QTR	Missing Expenditure /CU QTR
2 nd order	?	?	?
Mover-Stayer	?	?	?
Two part	na	\$5,182.59	\$3.46
Mod MS	< 0	-\$27,784.80	\$-18.52
<i>*Estimated from $P(A/W)$</i>			

Results, Reports / Expenditure

Music

Model	$P(A=1 / W=1)$	Missing Expenditure /QTR	Missing Expenditure /CU QTR
2 nd order	.445	\$58,551.30*	\$39.34*
Mover-Stayer	.741	\$16,409.11*	\$10.94*
Two part	na	-\$1,840.63	-\$1.23
Mod MS	.187	\$31,017.09	\$20.68
<i>*Estimated from $P(A/W)$</i>			

Are Models Worth It?₁

- Generally more information is better
- Time consuming – estimation is slow
- Model fit does suffer more than expected
- Are estimates of missing expenditure superior?
 - ▶ LCA Mover-Stayer, 2nd order, are vetted, stable over time, estimates make sense, internal validity, validation with external sources

Are Models Worth It?₂

- Estimates are no more believable for difficult expenditure categories (e.g. major appliances)
- Two part latent growth produces very different estimates
- Some support for modified mover-stayer
- Much more testing is needed
 - ▶ Grouping variables
 - ▶ Examine estimates over time
 - ▶ Validation with external sources

Contact Information

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Estimating Magnitude of Underreported Expenditures for False Negative: Notation

\hat{R}_c Total *reported* expenditures for persons with characteristics, c

$\hat{\pi}_{1|1,c}$ Accuracy rate for persons with characteristics, c , estimated from M-S model; i.e. $P(A=1|W=1)$

T_c True total expenditures persons with characteristics, c

$T_{c,+}$ True total expenditures persons with characteristics, c for true positives

$T_{c,-}$ True total expenditures persons with characteristics, c for false negatives

Assumptions

- No false positive reports of expenditures
- Reported expenditures are accurate; i.e.,

$$E(\hat{R}_c) = T_{c,+}$$

- Mean expenditures for reporters and mean expenditures for nonreporters are equal

Estimate of Underreports Due to False Negatives

Under these assumption, an estimate of T_c is

$$\hat{T}_c = \frac{\hat{R}_c}{\hat{\pi}_{1|1,c}}$$

Thus, an estimate of $T_{c,-}$ is

$$\hat{T}_{c,-} = \hat{T}_c - \hat{R}_c$$

Mover-Stayer Model

Assumptions

Population can be divided into:

- Persons who purchase the item in each quarter (“purchase-stayers”)
- Persons who do not purchase the item in any quarter (“nonpurchase-stayers”)
- Persons whose purchase behavior is not consistent across the quarters (“movers”)

Additional Assumption

- No false positive reports. Persons who report a purchase are assumed to have actually made that purchase.

Definition of Latent Variables

Where,

$$W = \begin{cases} 1, & \text{if one or more purchases of an item during the} \\ & \text{quarter ("purchaser")} \\ 2, & \text{if no purchase ("non-purchaser")} \end{cases}$$

with similar definition for X , Y , Z for 2nd, 3rd, and 4th interview

Definition of Indicator Variables

Define for Interview 1,

$A = 1$ if reported as a purchaser for the quarter
2 if reported as non-purchaser

with similar definition for B, C, D
for 2nd, 3rd, and 4th interviews

Grouping Variables

1. Family size
2. Refusal to answer income question
3. Derived variable combining records use and interview length
4. Income class