

Incorporating evidence on effect-heterogeneity in CEA

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THE CHOICE INSTITUTE

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Background

- > **Cost-effectiveness analysis is beginning to play a major role in decision-making for private and some public plans in the US**

CVS Caremark is initiating a program that allows clients to exclude any drug launched at a price of greater than \$100,000 per QALY from their plan. The QALY ratio is determined based on publicly available analyses from the Institute for Clinical and Economic Review (ICER), a non-profit organization dedicated to the development of cost-effective



The image is a screenshot of the Managed Healthcare Executive website. At the top, there is a search bar with the text "Enter your keywords". Below the search bar is a navigation menu with the following items: Topics, Industry Analysis, Business Strategy, Health Management, and Hospitals & Providers. The main content area features an advertisement for "SHED LIGHT ON INVOLUNTARY MOVEMENT DISORDERS" with the tagline "Complexity ... illuminated." Below the advertisement is an article titled "ICER-VA alliance for formulary management sparks debate" dated Nov 19, 2017. The article is categorized under "Benefit design and pricing, Business Strategy, Contracting, MHE Articles, Pharmacy, Pharmacy Best Practices, Pharmacy Strategy". At the bottom of the article, there are social media icons for Facebook, Twitter, LinkedIn, and Google+.



Criticisms

September 12, 2018

Mr. Larry J. Merlo
 President and Chief Executive Officer
 CVS Health
 One CVS Drive
 Woonsocket, Rhode Island 02895

Dear Mr. Merlo:

ACCSES
 Aimed Alliance
 Alliance for Aging Research
 Alliance for Patient Access
 American Academy of Nursing
 American Academy of Ophthalmology
 American Association of People with Disabilities
 Association of University Centers on Disabilities
 Autism Society of America
 Autistic Self Advocacy Network
 Bazelon Center for Mental Health Law
 Beyond Type 1
 Black Women's Health Imperative
 Bladder Cancer Advocacy Network
 Brain Injury Association of America
 California Consortium of Addiction Programs and Professionals
 Cancer Support Community
 CancerCare
 CARE About Fibroids
 Center for Autism and Related Disorders
 Center for Public Representation
 Cutaneous Lymphoma Foundation
 Davis Phinney Foundation
 Depression and Bipolar Support Alliance
 Diabetes Patient Advocacy Coalition
 Disability Rights Education and Defense Fund
 Epilepsy Association of North Carolina
 Epilepsy Foundation
 Epilepsy Foundation - Alabama
 Epilepsy Foundation Maryland
 Epilepsy Foundation Metropolitan Washington
 Epilepsy Foundation Nebraska
 Epilepsy Foundation New England
 Epilepsy Foundation Northwest
 Epilepsy Foundation of Arizona
 Epilepsy Foundation of Colorado
 Epilepsy Foundation of Connecticut

Epilepsy Foundation of Georgia
 Epilepsy Foundation of Greater Los Angeles
 Epilepsy Foundation of Greater Southern Illinois
 Epilepsy Foundation of Indiana
 Epilepsy Foundation of Iowa
 Epilepsy Foundation of Kentuckiana
 Epilepsy Foundation of Michigan
 Epilepsy Foundation of Middle and West Tennessee
 Epilepsy Foundation of Minnesota
 Epilepsy Foundation of Missouri and Kansas
 Epilepsy Foundation of Nevada
 Epilepsy Foundation of Northeastern New York, Inc.
 Epilepsy Foundation of Oklahoma
 Epilepsy Foundation of Vermont
 Epilepsy Foundation Ohio
 Epilepsy Foundation Utah
 Genetic Alliance
 Global Liver Institute
 Global Healthy Living Foundation
 Health Hats
 Illinois Association of Behavioral Health
 International Foundation for Autoimmune & Autoinflammatory Arthritis
 Judy Olsen
 Kidney Cancer Association
 Lung Cancer Alliance
 LUNGeVity Foundation
 Lupus and Allied Diseases Association, Inc.
 LymeDisease.org
 Men's Health Network
 Mended Hearts
 Mental Health America
 National Alliance on Mental Illness
 National Infusion Center Association
 National MPS Society
 National Multiple Sclerosis Society
 National Patient Advocate Foundation
 No Health without Mental Health
 Not Dead Yet
 Partnership to Improve Patient Care
 Patrick Gee
 Pediatric Congenital Heart Association
 PXE International
 RetireSafe

Rosie Bartel
 Roxanne Davenport
 TASH
 The AIDS Institute
 The Arc of the United States
 The Asthma and Allergy Foundation of America
 The diaTribe Foundation
 The National Council on Independent Living
 The Veterans Health Council
 Tuberos Sclerosis Alliance
 U.S. Pain Foundation
 United Cerebral Palsy
 United Spinal Association
 Vietnam Veterans of America

Therefore, we request that you reconsider this decision. CVS Health's stated purpose is "helping people on their path to better health." Reliance on cost-effectiveness thresholds like ICER's falls short of this purpose, replacing deeply personal, individual health care decisions with an opaque algorithm based on average study results that do not address the needs of different patients and special populations.



Premise

- > Relying on average cost-effectiveness of a new technology have been criticized in the presence of heterogeneity

- > Consider three issues in this talk:
 - Stochastic (first-order) uncertainty vs variability
 - Implication for learning by doing
 - Demand-weighted cost-effectiveness analysis
 - > Its relationship to indication-based pricing



Stochastic (first-order) uncertainty vs variability

- > Stochastic (first-order) uncertainty
 - Represents uncertainty in subject-level outcomes that is entirely due to chance.
 - E.g. even if you specify that subjects have a 5% chance of death, for any single individual at any point in time, either he dies and stay alive.
- > This uncertainty is due to pure randomness (e.g. flipping a coin) – UNPREDICTABLE
- > Cannot be used as a basis to allocate resources



Stochastic (first-order) uncertainty vs variability

> Variability

- PREDICTABLE differences in outcomes and costs for subgroups determined by subject characteristics

> Important for resource allocation

- Heterogeneity may also arise due to system characteristics and also individual preferences

> Efficient allocation of resources should try to directly incorporate variability in decision-making



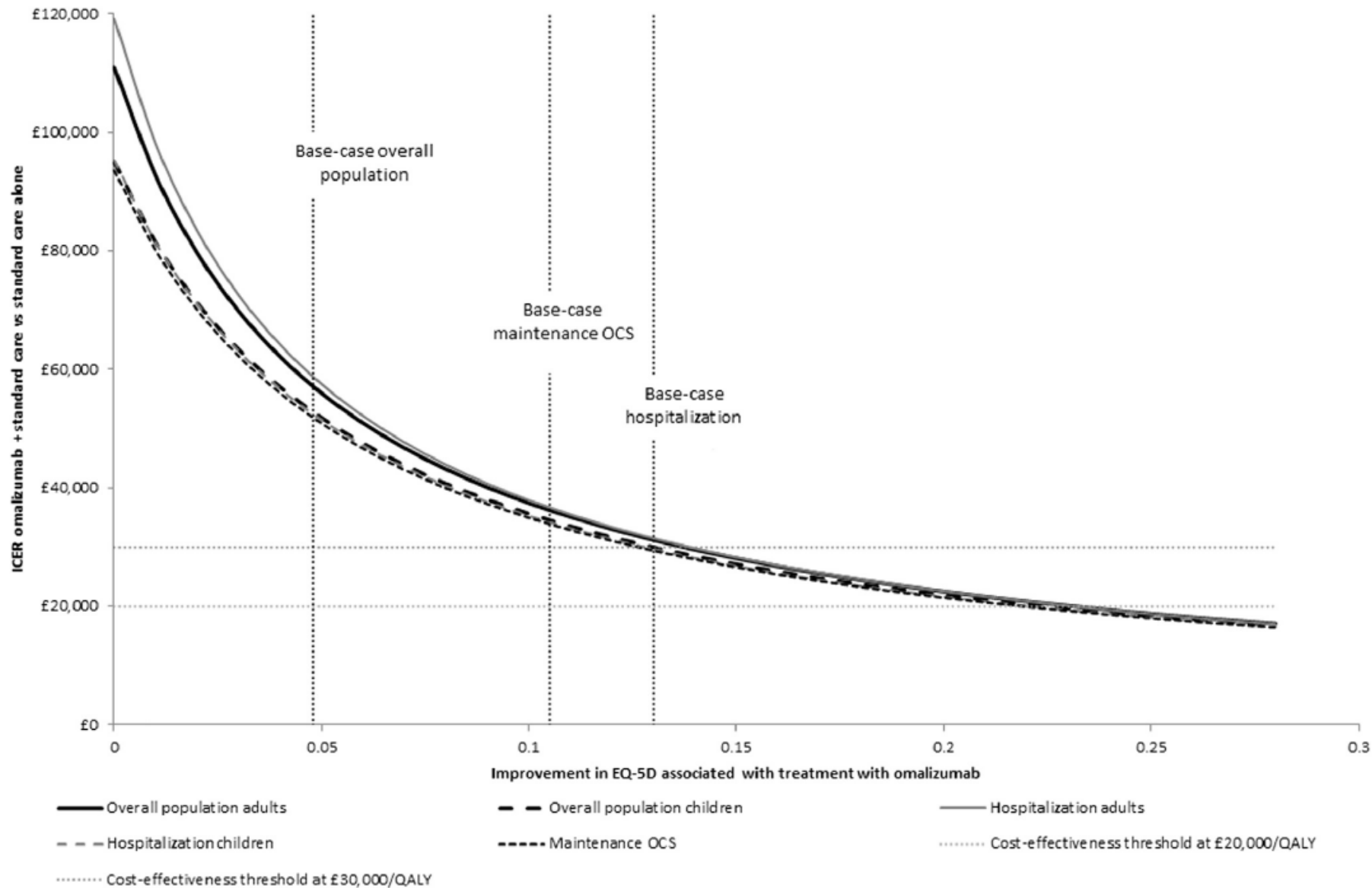


Fig. 2 – Effect of improvements in health-related quality of life on the cost-effectiveness of omalizumab. The subgroups with three or more exacerbations are not shown because the curves overlap the other subgroups and the overall population. EQ-5D, EuroQol five-dimensional questionnaire; ICER, incremental cost-effectiveness ratio; OCS, oral corticosteroid; QALY, quality-adjusted life-year.



Omalizumab for treating severe persistent allergic asthma

Technology appraisal guidance

Published: 24 April 2013

[nice.org.uk/guidance/ta278](https://www.nice.org.uk/guidance/ta278)

1 Guidance

1.1 Omalizumab is recommended as an option for treating severe persistent confirmed allergic IgE-mediated asthma as an add-on to optimised standard therapy in people aged 6 years and older:

- who need continuous or frequent treatment with oral corticosteroids (defined as 4 or more courses in the previous year), **and**
- only if the manufacturer makes omalizumab available with the discount agreed in the patient access scheme.

<https://www.nice.org.uk/guidance/ta278/>



Learning-by-Doing

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Learning-by-Doing: Transition of Stochastic Uncertainty to Variability

- > Physician and patients learn from the random variation in outcomes
- > Develop algorithms to identify subgroups with higher/lowered than average outcome
- > Should resource allocation be generous up-front to allow for learning?
 - Fundamentally a trade-off between current health & costs and future health
 - Depends on expected quality-of-learning and the rate of learning
 - Empirical evidence suggest that learning exists but far from perfect.



HETEROGENEITY IN ACTION: THE ROLE OF *PASSIVE* PERSONALIZATION IN COMPARATIVE EFFECTIVENESS RESEARCH

ANIRBAN BASU^{a,*}, ANUPAM B. JENA^b, DANA P. GOLDMAN^c, TOMAS J. PHILIPSON^d and ROBERT DUBOIS^e

^a*Department of Health Services, University of Washington, Seattle and the National Bureau of Economic Research, Cambridge MA, Seattle, WA, USA*

Table II. Predicted impact of generic group atypical antipsychotic drugs (AADs) compared with branded group AADs on average number of hospitalizations in 12 months following initiation of therapy

Group	All hospitalizations	Schizophrenia-related hospitalizations
	Mean (95% CI)	Mean (95% CI)
All patients (ATE)	0.35 (0.02, 0.67)	−0.07 (−0.28, 0.10)
Patients initiating therapy with generic group (TT)	0.17 (−0.17, 0.44)	−0.15 (−0.38, −0.03)
Patients initiating therapy with branded group (TUT)	0.61 (0.29, 1.05)	0.002 (−0.13, 0.22)
TT—ATE	−0.18 (−0.13, −0.28)	−0.08 (−0.04, −0.12)

ATE, average treatment effect; TT, effect on the treated; TUT, effect on the untreated.



MECHANISMS TO REDUCE OPTIMAL DURATION OF OUTCOMES-BASED CONTRACTING

ANIRBAN BASU, JOSH J CARLSON
University of Washington, Seattle
uwchoice.org

Structured learning-by doing

- > Centralized learning from doing
 - Learning during the outcomes-based agreement
 - Tuesday Poster 3:00 – 7:00 pm

- > Needs structure
 - Determine time for learning
 - Select methods to learn faster
 - Have explicit decision-making tied to the end of learning period

BACKGROUND

OBJECTIVES: For a payer's perspective, entering an outcomes-based contract (OBC) can reduce risk in the face of uncertainty about treatment effectiveness. However, implementing OBC is often costly, and sustaining it may be even costlier. Little is known about the optimal duration of OBC.

METHODS: A central premise for terminating an OBC is whether uncertainty about the treatment effectiveness is resolved. The optimal length of an OBC can be reduced if payers are willing to learn from their own data during the execution of an OBC contract. We combine concepts from Value of Information methods and Bayesian designs to compare alternative designs of learning within the real-world system. Especially we compare 1) static approaches, where sample size are determined a priori, either through traditional sample size calculation or Expected Value of Sample Information (EVS) criterion, and 2) dynamic approaches, that follow adaptive allocation of patients to treatments following evolving posterior means or posterior variance. Using simulated scenarios, we compare the time to resolution of uncertainty under each of these mechanisms with 1000 Monte-Carlo runs.

METHODS

Expected Value of Monitoring

$$V^* = \sum_{i=1}^L \beta^{i-1} \cdot i \cdot ((1-p_i) \cdot \lambda_A + p_i \cdot \lambda_B) \cdot D - D^*$$

i = Total incident patient population every year = 2400
 p_i = The proportion of the patient population using treatment B in any period. Initially, under equal coverage of A and B, it is assumed to be 25%. If CER resolves uncertainty and identifies A to be superior, then p_i is assumed to change to 100%. If B is identified as superior, p_i changes to zero. Such a decision change is assumed to happen at the end of the CER trial.
 L = length of the trial in years rounded to nearest integer = $\text{round}(n_{\text{total}} / (20 \cdot 12))$.
 D = Monetized value of a life year. Assumed to be \$100,000/year.
 D^* = Cost of monitoring in the real-world setting.

Comparative Designs



Data Generating Processes

Scenario	TRUE INFORMATION		PREDICTABLE INFORMATION	
	Drug A P ₁	Drug B P ₂	U ₁ (A, B)	U ₂ (A, B)
DGP1	0.5	0.5	1.0, 0.0	0.0, 1.0
DGP2	0.5	0.5	1.0, 0.0	0.0, 1.0
DGP3	0.5	0.5	1.0, 0.0	0.0, 1.0

QUICK FACTS & FINDINGS

Fundamental premise: OBC stopped when decisional uncertainty is resolved.

Framework: Value of information to decide the optimal duration of an OBC using real-time monitoring of data.

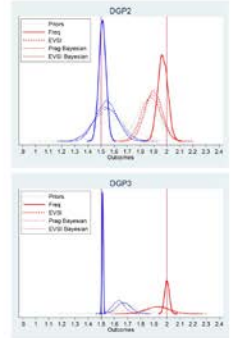
What we study: alternative allocation rules that guide how a payer may manipulate access to new versus older drugs during the OBC period to generate the required evidence in the least amount of time.

Methodology: sample size calculation, expected value of sample information and the Bayesian adaptive allocation.

Main result: The average time to uncertainty resolution was 1/20 to 1/70 for dynamic approaches compared to static approaches.

Implications: Payers can shorten time to termination of OBC by engaging in active learning during the OBC period.

Payers may want to steer patients to use the treatment with the greatest uncertainty in its effectiveness during the OBC period.



MAIN RESULTS



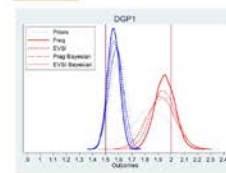
RESULTS SUMMARY

The average time to uncertainty resolution was 1/20 to 1/70 for dynamic approaches compared to static approaches. Among the static approaches, EVSI approach required a shorter time. Under the dynamic approaches, adaptive allocation based on posterior variance had the shortest time to uncertainty resolution. Dynamic allocation of patients to alternative treatments are found to be superior approaches as they help resolve decisional uncertainty faster than the uncertainty in the true comparative effect.

IMPLICATIONS

Payers can shorten time to termination of OBC by engaging in active learning during the OBC period. Payers may want to steer patients to use the treatment that has the greatest uncertainty associated with its effectiveness during the OBC period. Further work requiring the application of methods to specific clinical scenarios will be most useful.

ACCUMULATED INFORMATION



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 Josh Carlson: Carlsoj@uw.edu

This research was supported through unrestricted funds from a consortium of twelve biomedical companies to the University of Washington, Seattle.

Demand-weighted CEA

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Demand-weighted cost-effectiveness

- > Evidence on variability is important even if
 - there is no learning-by-doing
 - there is no opportunity to implement sub-group-based coverage

ORIGINAL ARTICLE

New Metrics for Economic Evaluation in the Presence of Heterogeneity: Focusing on Evaluating Policy Alternatives Rather than Treatment Alternatives

*David D. Kim, PhD, Anirban Basu, PhD
Medical Decision Making 2017*



ICER for Alternative Treatments

Typical ICER compares Treatment A vs B

$$\begin{aligned} ICER &= \frac{E(C_A) - E(C_B)}{E(Q_A) - E(Q_B)} = \frac{E(\Delta C_{AB})}{E(\Delta Q_{AB})} \\ &= \frac{\sum_j \{P_j \cdot E(\Delta C_{AB,j})\}}{\sum_j \{P_j \cdot E(\Delta Q_{AB,j})\}} \quad j = 1, 2, 3 \end{aligned}$$

P_j = Size of Subgroup j

- Suppose, clear evidence on variability in ICER across subgroups.



ICER for Alternative Treatments

- ICER comparing potential realized value of Treatment A vs B

$$ICER = \frac{\sum_j \{P_j \cdot D_j \cdot E(\Delta C_{AB,j})\}}{\sum_j \{P_j \cdot D_j \cdot E(\Delta Q_{AB,j})\}} \quad (4)$$

- D_j : the rate of adoption of treatment A in the population subgroup j
- Similar to “Volume weighted price” across indications



ICER for Alternative Policies

- The Rate of the uptake is endogenous to the policy
- So ICER should be comparing two coverage policies – Policy k vs Status quo

$$ICER_{policy_k} = \frac{\sum_j \{P_j \cdot D_{jk}(f_k) \cdot E(\Delta C_{AB,j}) \cdot f_k\}}{\sum_j \{P_j \cdot D_{jk}(f_k) \cdot E(\Delta Q_{AB,j})\}}$$

- f_k : a fraction of the incremental costs under a policy k borne by a payer



Table 1 Illustration of Traditional and Modified ICERs and INMBs under a Health-Care Sector

Parameter	Males	Females	Overall
Total costs per patient under Statin + Fibrate, \$	\$107,021	\$107,023	-
Total costs per patient under Statin Only, \$	\$98,131	\$98,131	-
Total incremental costs per patient, \$	\$8,890	\$8,892	-
Total QALYs per patient under Statin + Fibrate, \$	9.468	9.308	-
Total QALYs per patient under Statin Only, \$	9.200	9.200	-
Total incremental QALYs per patient, \$	0.268	0.108	-
Subgroup-specific ICER	\$33,130/QALY	\$82,562/QALY	-
Subgroup-specific INMB ^a	\$3,170	-\$4,032	
Subgroup size (P_j)	0.533	0.467	
Traditional population ICER (eq. 3)			\$46,000/QALY
Population NMB per patient from statin monotherapy ^b			\$315,869
Adoption of Statin + Fibrate under status quo ($f_k = 0.80$), D_j	0.072	0.043	
Modified population ICER ($f_k = 0.80$, eq. 7)			\$41,733/QALY
Status-quo policy NMB per patient ($f_k = 0.80$, eq. 8)			\$315,910
Adoption of Statin + Fibrate under Policy 2 ($f_k = 1.0$), D_j^c	0.075	0.045	
Modified population ICER under Policy 2 ($f_k = 1.0$, eq. 7)			\$41,766/QALY
Policy 2 NMB per patient ($f_k = 1.0$, eq. 8)			\$315,911
Adoption of Statin + Fibrate under Hypothetical Policy, ^d D_j	0.23	0.023	
Population ICER under Hypothetical Policy ^d (eq. 7)			\$34,848/QALY
Hypothetical Policy ^d NMB per patient (eq. 8)			\$316,214

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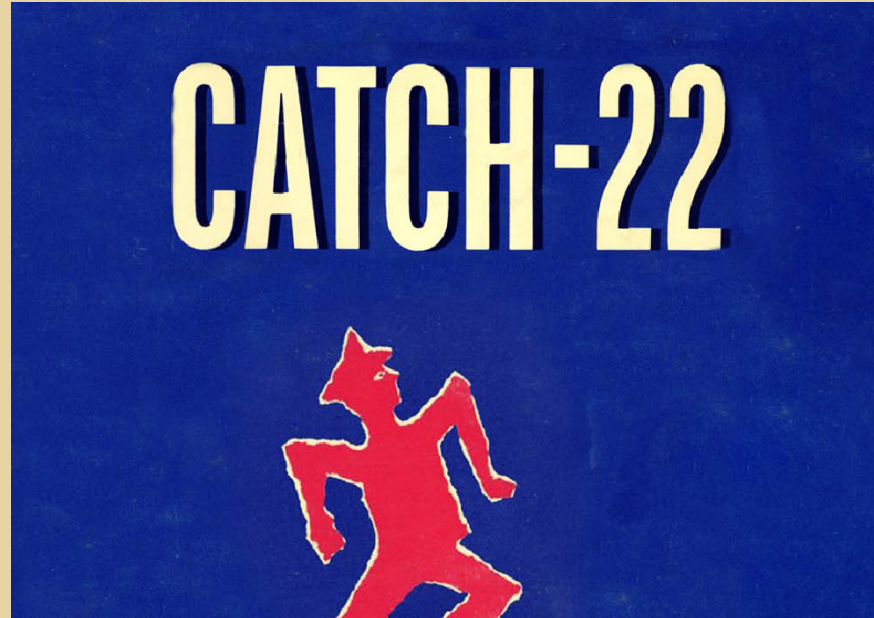
Future Work in CEA

- To use demand weighted CEA at launch
 - Can we develop reliable estimates for *evidence elastic of demand*?
 - *Discrete choice experiments*
 - *Retrospective analysis*
 - Validated prediction model for technology diffusion
- CEA at 5 year assessment
 - Direct estimate from real-world use.



Crossroads

Decision makers
focus of
population
averages because
of the lack of
reliable evidence
on heterogeneity



Manufacturers argue
that there is no
incentive to generate
evidence on
heterogeneity



Conclusions

- **Are we are failing to produce the necessary evidence of heterogeneity of effects, which can improve value in the society, by not providing sufficient reward incentive for such information.**
- > **Importantly to create an environment that respects and rewards evidence on heterogeneity.**
- > **Laying a clear path of incorporating reliable evidence on heterogeneity in third-party assessor's base analysis. This includes**
 - **not reporting population average cost-effectiveness results when there are distinct differences in subgroup-specific results,**
 - **experimenting with demand weighted cost-effectiveness approaches.**

