

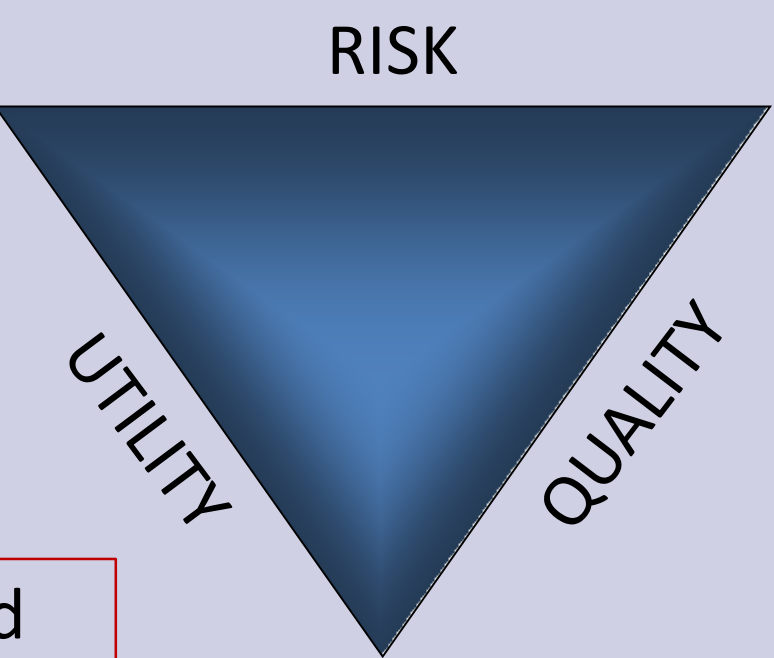
REDUCING DIAGNOSTIC ERROR IN MEDICINE

BY BREAKING THE DIAGNOSTIC AND EPIDEMIOLOGIC FEED FORWARD LOOP



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1. Describe how diagnostic tests may influence the prevalence of a disease. 2. Explain how, by increasing apparent prevalence, a test may influence apparent its apparent diagnostic power. 3. Show how the cycle of increasing test sensitivity and screening leads to an increase in Diagnostic Error in Medicine.

SITUATION

The power of screening and diagnostic tests is dependent on factors that influence the maximum sensitivity and specificity that can be achieved. However, it is the prevalence of a disease that determines the Positive Predictive Value [PPV] and Negative Predictive Value [NPV] of a test, and so, the total number of false positive and false negative diagnoses.

Unfortunately, epidemiologic data used to estimate prevalence is often based on the results of the test itself. And this data influences clinical judgment, societal behavior, and public policy. In turn, this leads to a paradox whereby the test initiates a feed forward cycle of increased perceived reliability, increased use, and, falsely increased prevalence.

PROBLEM

How can we

RISK	Maximize patient safety through collection of accurate and precise epidemiologic data regarding test reliability
QUALITY	Minimize discomfort and the pain suffered due to over diagnosis and treatment due to wrong epidemiologic data
UTILITY	Minimize expenditure of scarce resources by improving the reliability of epidemiologic data based on diagnostic tests

SOLUTION

It is proposed that the present feed forward cycle can be broken through the establishment of **epidemiologic trials**. These would be designed along the lines of clinical trials whereby the clinical test modality and the diagnostic decision making process by which the modality is utilized are evaluated prior to general deployment through statistical modeling and by assessing clinical outcomes in an adequate number of patients.

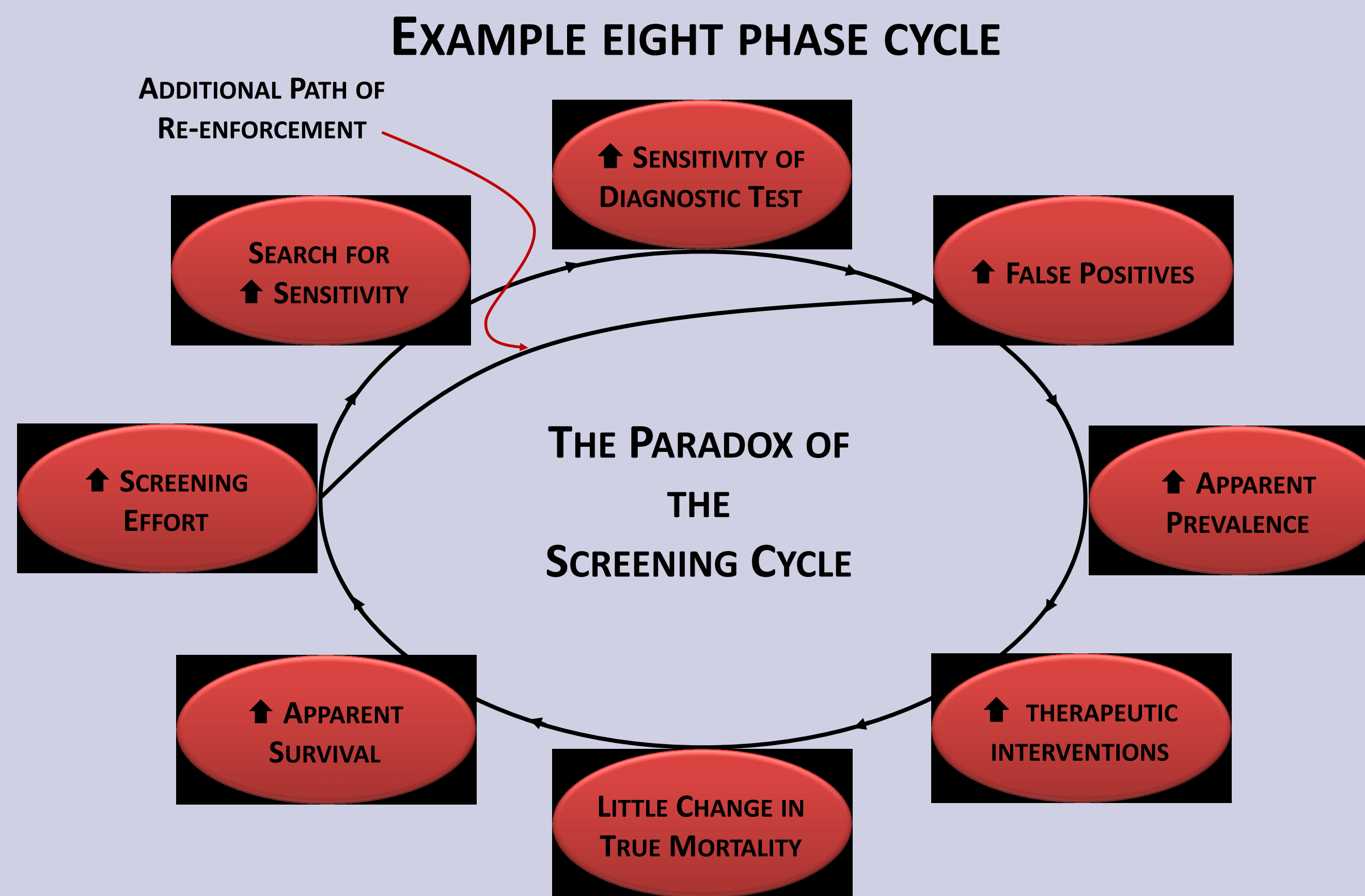
This approach will, at the very least, delay the general implementation of a flawed clinical testing modality or diagnostic decision making process thereby avoiding the corruption of epidemiologic data used to develop clinical guidelines, influence societal attitudes, and formulate public policy.

IMPLEMENTATION

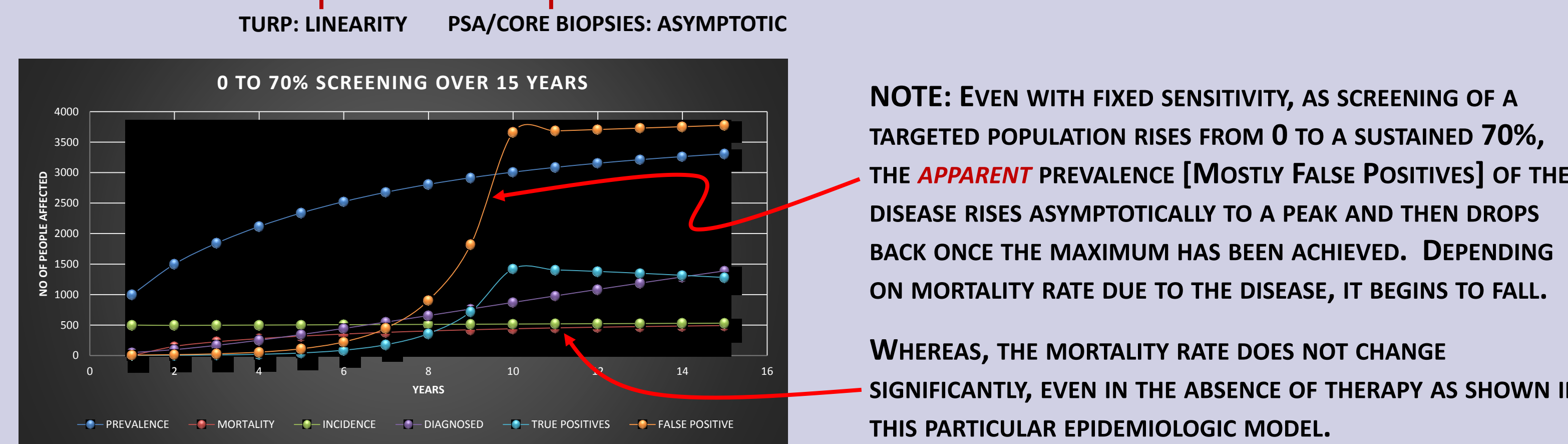
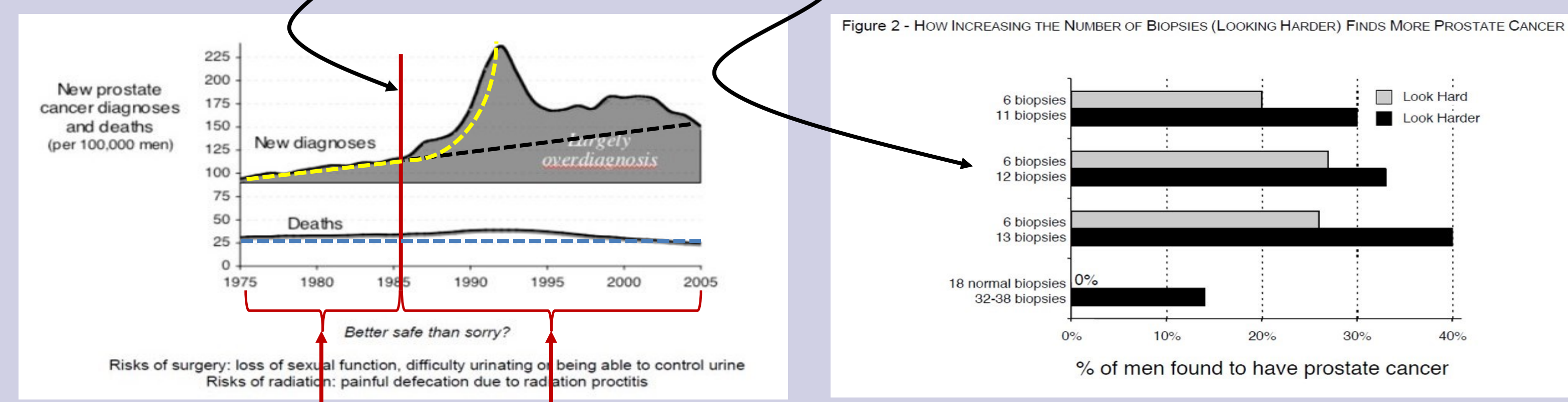
A literature search was carried out to identify diseases where prevalence has risen while the death rate has stayed the same or dropped slightly. These were studied to determine how test characteristics might have influenced reported disease prevalence. Several were identified including breast cancer, thyroid cancer, prostate cancer, and malignant melanoma.

An MS Excel spreadsheet was developed to model the impact rising rates of screening can have on apparent prevalence through rising false positive diagnoses. The spreadsheet data was then plotted and compared to epidemiologic data of actual screening efforts; in this case prostate cancer. Changing rates of sensitivity and mortality were excluded for simplicity.

THE EPIDEMIOLOGIC FEED FORWARD LOOP



↑ SCREENING + [↑ SENSITIVITY ⇒ ↓ SPECIFICITY] ⇒ FALSELY ↑ APPARENT PREVALENCE



SIMPLIFIED EPIDEMIOLOGIC MODEL OF A POPULATION AFFLECTED BY ONLY ONE DISEASE WITH NO THERAPEUTIC INTERVENTION

POPULATION	POP GROWTH	INCIDENCE	PREVALENCE	MORTALITY	% DIAGNOSIS	FINAL SCREEN	SENSITIVITY	SPECIFICITY								
100000	0.01	0.005	0.01	0.15	0.05	70.00%	0.95	0.95								
YEAR	POPULATION	INCIDENCE	PREVALENCE	MORTALITY	% DIAGNOSIS	ACCEPTANCE	SCREENED	TRUE +	TRUE -	FALSE +	FALSE -	TOTAL +	TOTAL -	PPV	NPV	F1-T1
1	100000	500	1000	0	10	5.00%	137	1	129	7	0	58	129	2.13%	99.95%	5.40
2	101000	495	1500	150	98	6.50%	276	4	259	14	0	115	259	3.17%	99.93%	3.74
3	101860	498	1845	225	168	9.09%	556	9	520	27	0	204	521	4.28%	99.91%	3.14
4	102654	500	2118	277	251	11.88%	1120	19	1047	55	1	326	1048	5.95%	99.90%	2.84
5	103403	503	2341	318	345	14.73%	2254	41	2107	111	2	497	2110	8.34%	99.90%	2.67
6	104120	505	2526	351	445	17.60%	4536	87	4241	223	5	754	4246	11.47%	99.89%	2.58
7	104810	508	2680	379	549	20.47%	8123	177	8535	449	9	1175	8544	15.08%	99.89%	2.54
8	105479	511	2809	402	655	23.32%	18344	358	17178	904	19	1917	17197	18.68%	99.89%	2.52
9	106132	513	2918	421	763	26.15%	36879	716	34572	1820	38	3299	34610	21.72%	99.89%	2.54
10	106772	516	3010	438	871	28.93%	74131	1423	69581	3662	75	5955	69656	23.89%	99.89%	2.57
11	107402	519	3088	451	978	31.66%	14497	2804	14009	6885	144	12054	14093	26.14%	99.89%	2.63
12	108024	522	3156	463	1083	34.33%	29189	5478	28458	13708	284	25516	28531	28.34%	99.90%	2.69
13	108641	524	3214	473	1187	36.99%	59218	10818	58099	2732	531	62666	58630	30.11%	99.90%	2.77
14	109255	527	3265	482	1288	39.45%	12576	2115	12340	3755	1001	13347	12440	32.08%	99.90%	2.86
15	109865	530	3310	490	1387	41.90%	26535	4279	22181	3778	1977	28162	22158	33.91%	99.91%	2.95

REFERENCES – SELECTED:
 1. Feinstein AR; Clinical Epidemiology: Parts I, II, and III; Annals of Internal Medicine Vol 69: No's 4-6, Oct – Dec 1968.
 2. Feinstein AR; Fraud, Distortion, Deception, and Consensus: The Problems of Human and Natural Deception in Epidemiologic Science; The American Journal of Medicine Vol 84 Mar 1988 p 475-478.
 3. Feinstein AR; Double Standards, Scientific Methods, and Epidemiologic Research; New England Journal of Medicine Vol 307 No 26 Dec 1982 p 1611 – 1617.
 4. Harach RH, Franssila KO, Wassenius VM; Occult Papillary Carcinoma of the Thyroid: A "Normal" Finding in Finland. A Systematic Autopsy Study; Cancer Vol 56 No 3 Aug 1985 p 531 – 538.
 5. Crosswell JM; Cumulative Incidence of False-Positive Results in Repeated, Multimodal Cancer Screening; Annals of Family Medicine Vol 7 No 3 May/June 2005 p 212 – 222.
 6. Day NE; Quantitative Approaches to the Evaluation of Screening Programs; World Journal of Surgery Vol 13 No 1 Jan/Feb 1989 p 3 – 8.
 7. Cameron HM, McGoogan E; A Prospective Study of 1152 Hospital Autopsies: Inaccuracies in Death Certification; Journal of Pathology Vol 133 1981 p 273 – 283.
 8. Skendzel LP; How Physicians Use Laboratory Tests; The Journal of the American Medical Association Vol 239 No 11 Mar 1978 p 1077 – 1080.
 9. Feinstein AR, Esdaille JM; Incidence, Prevalence, and Evidence: Scientific Problems in Epidemiologic Statistics for the Occurrence of Cancer; The American Journal of Medicine Vol 82 Jan 87 p 113-23.
 10. Knapp RG, Miller MC; Clinical Epidemiology and Biostatistics; Williams and Wilkins 1992. Chapters 3 – 4 p 31 - 60.

COST BENEFIT ANALYSIS

Presently, the massive growth in cost of healthcare can be traced to a number of issues that have influenced the way screening and diagnostic medicine is pursued.

One of these is the rise in the costs incurred by screening for clinically occult disease. These costs include increased monetary outlay due to over utilization and decreased patient safety due to adverse effects of the diagnostic process combined with overtreatment.

What is not shown in this poster is the effect of drift in diagnostic criteria that tend to cause a rise in false positive diagnoses due to feed back of societal and medicolegal pressures to avoid missing a diagnosis.

EXAMPLE

There has been a significant rise in the diagnosis of thyroid, breast, and prostate cancer without a significant rise in mortality.

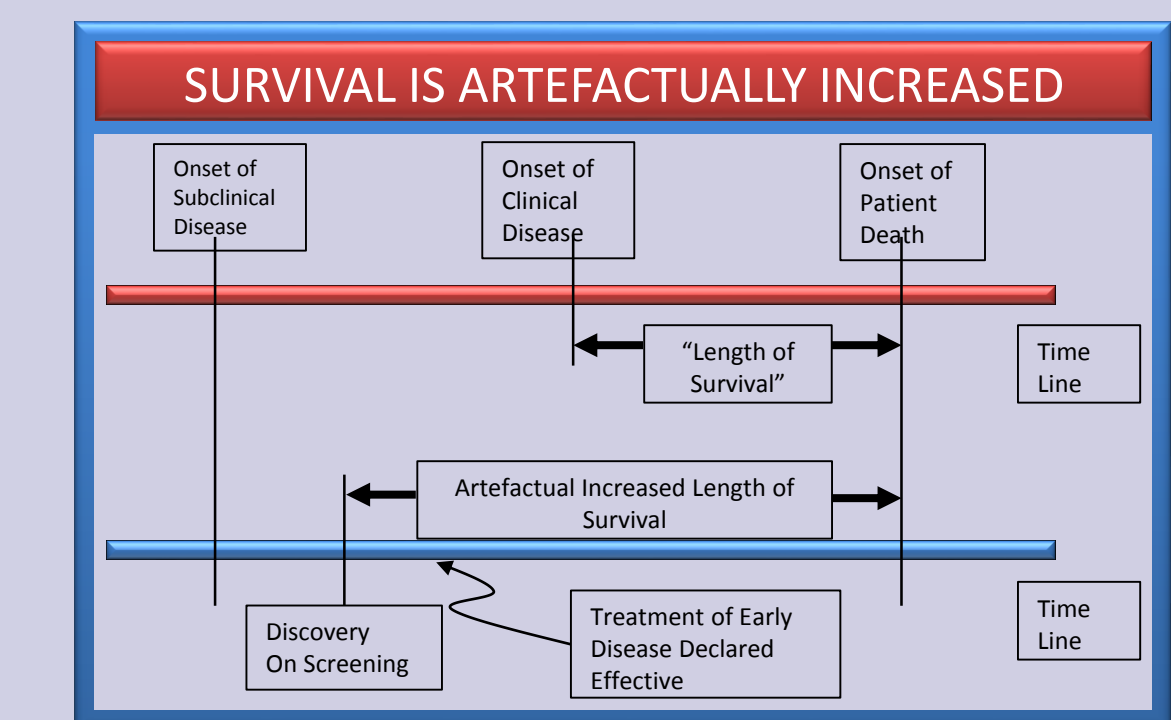
In this poster the effect of increased screening and increased sensitivity due to increased numbers of prostate core biopsies taken are illustrated. For any methodology this reduces specificity increasing false positives.

It should be noted that part of the rise in diagnoses is due to rising median age of the baby boomer population while the drop of mortality back to the baseline may be due to new therapies that affect advanced stage disease.

However, the model clearly shows test characteristics and utilization patterns influence apparent disease prevalence and, so, perceived screening/diagnostic power of a test. In addition, screening leads to:

- ➔ Earlier identification of True Positives artificially extending survival
- ➔ Earlier treatment of True Positives possibly actually extending survival

These two facts confound epidemiologic data used to make clinical and policy decisions.



CONCLUSION

The complex interaction of various clinical and societal factors greatly influences the rate of diagnostic error and so, epidemiologic error.

As the test utilization and/or test sensitivity rises the specificity of a test falls increasing the number of false positives leading to an asymptotic [not asymptomatic!] rise in apparent prevalence.

This resulting rise in prevalence, due predominantly to rising false positives, increases apparent positive predictive value of the test further overstating the power of the test in question leading to increased:

DIAGNOSTIC ERROR IN MEDICINE