Best Practices in Oncology The Medical Necessity of Diagnostic Imaging and Testing in Medical Oncology March 28, 2019

"Oncology High-Value Best Practices" Webinar Series, Webinar #4

CALIFORNIA QUALITY COLLABORATIVE Breakthroughs for Better Health Care

Tech Tips – Zoom Meetings

Attendees are automatically MUTED upon entry	Refrain from using the hold button
Use the chat box, raise your hand, or <i>unmute</i> <i>yourself and jump in</i> if you have questions or would like to participate	Direct messages to Jose if you have any technical issues



End Meetin







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CALIFORNIA QUALITY COLLABORATIVE Breakthroughs for Better Health Care

Today's Speakers



- Bart Wald, MD
- Medical Director, California Quality Collaborative



• Eric Chevlen, MD



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Who is the California Quality Collaborative (CQC)?

CQC is a health care improvement organization dedicated to advancing the quality and efficiency of the health care delivery system in California. CQC creates scalable, measurable improvement in the care delivery system important to patients, purchasers, providers, and health plans.

- Started in 2007
- Multi-stakeholder governance
 - Core funding from health plans sharing a delivery system
 - Administered by the Pacific Business Group on Health
- **Purpose:** Identify and spread best practices across outpatient delivery system in California
 - Trains 2,000 individuals from 250 organizations each year









Oncology Series Webinar Dates





- Benefits & Limitations of Oncology Guidelines (Anthony Ciarolla, MD)
- Personalized Medicine (Mark Pegram, MD)
- Palliative Care

(Kavitha Ramchandran, MD)

The Medical Necessity

 of Diagnostic Imaging
 and Testing in Medical
 Oncology
 (Eric Chevlen MD)

"Is this test really necessary, doctor?"

UNDERSTANDING MEDICAL NECESSITY OF DIAGNOSTIC IMAGING AND LABORATORY TESTING IN MEDICAL ONCOLOGY

Eric Chevlen, MD

Lecture Outline

- Principles of medical necessity
- Types of testing
- Principles of screening tests (Bayes' theorem)
- Treatment-guiding tests
- Prognostic tests
- Surveillance tests
- Conclusions

Benefit of health plan medical necessity policies

- Evidence-based
- Reduces variation in decision by one reviewer
- Reduces variation in decision between reviewers
- Part of contract between health plan and member

Types of tests

Test type	Definition	Example
Screening	No signs or symptoms of disease	Mammography
Diagnostic	Signs or symptoms, but no confirmed diagnosis	Chest X-ray for coughing smoker
Staging	Confirmed diagnosis, extent of disease unknown	PET scan for clinical stage I lung cancer
Treatment- guiding	Diagnosis and extent of disease known, ideal treatment not known	HER2 assay in breast cancer
Surveillance	Completed treatment, no signs or symptoms	CT scan after treatment of small cell lung cancer
Prognostic	Disease and stage known, likely outcome unknown	Genetic assay of untreated prostate cancer

Tacit assumptions of screening

- Some diseases can be detected before they cause symptoms
- Early detection improves health outcomes.
- Examples:
 - Hypertension
 - Hypercholesterolemia
 - Hypothyroidism in newborns
- ► Counter-examples:
 - Shingles (cannot be detected before symptomatic)
 - Alzheimer's disease (early detection not shown to improve health outcomes)

Screening for cancer

Three patterns of cancer:

- Early and rapid dissemination to metastatic sites
 - Ovarian cancer
 - Plasmacytoma / myeloma
- Slow local progression without early metastasis
 - ► Well-differentiated prostate cancer in the elderly
- Early asymptomatic period, during which disease is detectable and curable, followed by incurable metastases
 - Breast cancer
 - Cervical cancer

Screening for cancer

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Three patterns of cancer



False positives and false negatives



SENSITIVITY

Sensitivity: if the disease is present, the test is positive. Sensitivity = inverse of false negative rate. High sensitivity = low false negative rate.

100 %	00 %	Disease status	X	X	X	x	X	X	ο	0	ο	ο
S	Sensitivity	Test result	POS	POS	POS	POS	POS	POS	NEG	NEG	NEG	NEG

X = disease present

67%	Disease status	X	X	X	x	x	X	ο	ο	ο	ο
Sensitivity	Test result	POS	POS	POS	POS	NEG	NEG	NEG	NEG	NEG	NEG

O = disease absent

Also 100 %	Disease status	x	x	X	X	X	X	ο	ο	ο	ο
Sensitivity	Test result	POS									

SPECIFICITY

Specificity: If the disease is absent, the test is negative. Specificity = inverse of false positive rate. High specificity = low false positive rate..

100 %	Disease status	X	X	X	X	X	X	ο	ο	0	ο
Specificity	Test result	POS	POS	POS	POS	POS	POS	NEG	NEG	NEG	NEG

X = disease present

75%	Disease status	X	X	x	X	x	X	0	ο	ο	ο
эреспісну	Test result	POS	POS	POS	POS	POS	POS	NEG	NEG	NEG	POS

O = disease absent

Also 100 %	Disease status	x	X	X	X	X	X	ο	ο	Ο	ο
Specificity	Test result	NEG									

Sensitivity/selectivity vs predictive value

In assessing sensitivity and selectivity, we KNOW whether the condition is present, and we ASK whether the test result corresponds with that known condition.

In assessing predictive value of a test, we KNOW the test result, and ASK whether the presence of the condition corresponds with that known test result.

	WE KNOW	WE ASK		
Sensitivity/selectivity	Presence of condition	Test results		
Predictive value	Test results	Presence of condition		

Predictive value

- Positive predictive value: if the test result is positive, the patient has the condition.
- Negative predictive value: if the test result is negative, the patient does not have the condition.
- Predictive values depend on:
 - ► Test sensitivity
 - Test specificity
 - Prevalence of condition in tested population

Calculating predictive value

Positive predictive value =

Negative predictive value =

true positives true positives + false positives true negatives

true negatives + false negatives

Importance of prevalence in testing -1

Test 1000 people Assume prevalence of condition in tested population = 10% Assume test sensitivity of 90% Assume test specificity of 90%

Test result	Condition present	Condition absent	Total
Positive	90 (true positive)	90 (false positive)	180 positives
Negative	10 (false negative)	810 (true negative)	820 negatives
Total	100	900	1000

Positive predictive value = true positives/true + false positives. Positive predictive value = 90/180 = 50% Negative predictive value = true negatives/true + false negatives. Negative predictive value = 810/820 = 98%

Importance of prevalence in testing - 2

Test 1000 people Assume prevalence of condition in tested population = 5% Assume sensitivity of 90% Assume specificity of 90%

Test result	Condition present	Condition absent	Total
Positive	45 (true positive)	95 (false positive)	140 positives
Negative	5 (false negative)	855 (true negative)	860 negatives
Total	50	950	1000

Positive predictive value = true positives/true + false positives. Positive predictive value = 45/140 = 32% Negative predictive value = true negatives/true + false negatives. Negative predictive value = 855/860 = 99.4%

Screening for breast cancer



Challenges of screening:

Prevalence in young women is low. Radiation causes cancer. Old women have competing causes of death. Relative risk (RR) reduction of ~ 15%. Absolute mortality risk reduction ~ 40 per 100,000.

JAMA, vol. 314(15), p.1615., 2015.

Screening standard risk subjects for lung cancer by CXR does not improve lung cancer death rate.

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Cumulative person-vears

- 155,000 prospectively ٠ randomized
- Standard risk subjects ٠
- Screening by chest radiograph annually x 4 years



77 286 154 116 230 348 305 902 380 725 454 719 527 804 599 790 669 955 734 523 788 854 831 678 863 330

The Prostate, Lung, Colorectal, and Ovarian (PLCO) Randomized Trial JAMA. 2011;306(17):1865-1873.



Years since Randomization

screening. N Engl J Med. 2011 Aug 4;365(5):395-409.

Staging of cancer with PET

If prevalence of <u>metastatic</u> disease at diagnosis is low, false positives far exceed true positives.

Example: early stage breast cancer

If metastatic disease is known from other studies, finding more metastases via PET does not improve outcome. Example: stage IV colon cancer.

For many cancers, re-assessment via PET is no better than reassessment via CT scan.

Risk prediction vs. risk reduction

Predicting increased risk does not improve outcome if no risk-reducing intervention is available.

Example: Decision-DX for uveal melanoma.

Predicting increased risk does not improve outcome if it dictates no change in therapy.

Example: Factor V Leiden in patient with recurrent deep vein thromboses.

Surveillance imaging may not improve outcome: Hodgkin lymphoma in first remission

	Number of patients	Number relapsing	Relapsing at 2 years or more	Relapse rate	5-year overall survival	Diagnostic images per detected relapse
Routine clinical follow-up + routine imaging	305	28 (9%)	4	13%	94%	47.5
Routine clinical follow-up + imaging only in case of relapse suspicion.	63	8 (13%)	2	9% Br J Haema	94% tol. 2014 Mar;16	4.7 4(5):694-700.

Molecular (genomic) profiling of cancer

Examples: FoundationOne, Guardian, Colaris

Tests for mutations of scores to hundreds of genes

A few tested genes may predict response to therapy, e.g. EGFR in lung cancer.

Specific predictive tests are available outside of a panel.

No evidence that therapy chosen on basis of test panel improves outcomes...

...yet.

Conclusions

Not all good ideas are proven ideas.

Evidence-based medicine requires...

...evidence!

Unnecessary testing is always expensive, and often harmful.