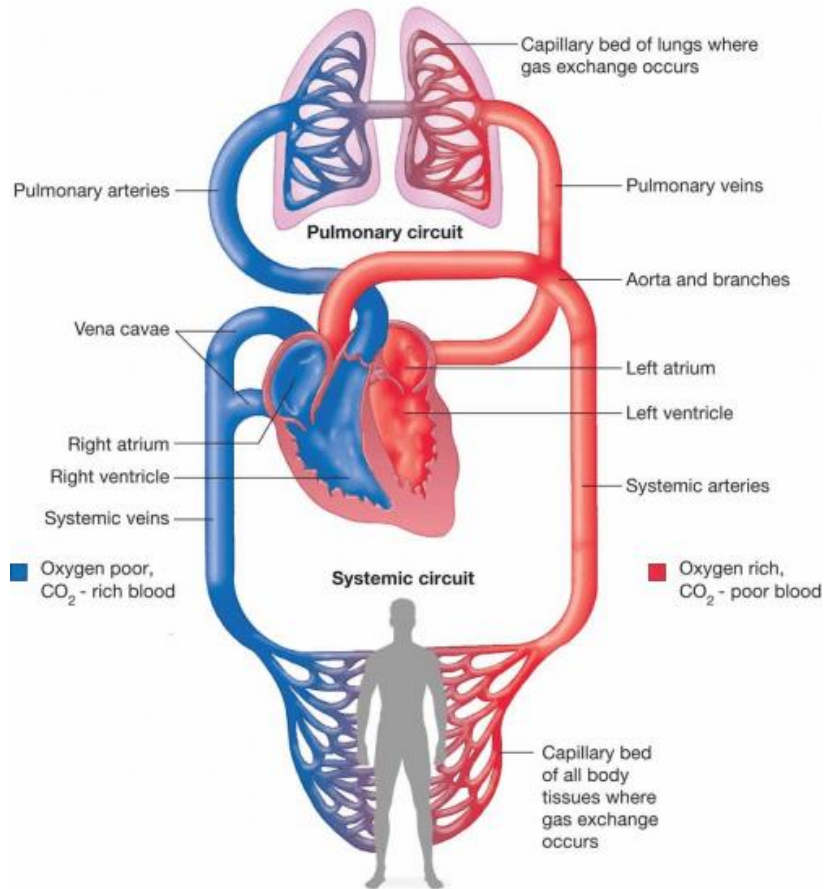


What makes something clinically useful?

From ROC curves to clinical practice

Rens ter Maat, MD, PhD candidate

Pulmonary embolism



Pathophysiology

- A thrombus (blood clot) in the venous system
- This clot then travels to the pulmonary capillary bed

Symptoms

- Swollen leg
- Shortness of breath
- Cardiac arrest

Diagnosis

- Chest-CT with IV contrast

Treatment

- Anticoagulants



Patient Alfred

Patient

- Male
- 55 years old
- Present at the emergency department with shortness of breath

Prior history

- Has atrial fibrillation (common heart rhythm abnormality)
- Takes anticoagulation medication

Signs and symptoms

- Shortness of breath since this morning
- Left leg hurts

Physical examination

- Breathing 22/min
- Saturation of 97%
- Swollen left leg



If a test does not change decisions, it is not useful.



Patient Dana

Patient

- Female
- 55 years old
- Present at the emergency department with shortness of breath

Prior history

- Had knee surgery 2 weeks ago
- Takes no medication

Signs and symptoms

- Shortness of breath since three days
- No other complaints

Physical examination

- Breathing 25/min
- Saturation of 94%
- Heart rate of 120/min



Making a biomarker for predicting pulmonary embolism

Biomarker A

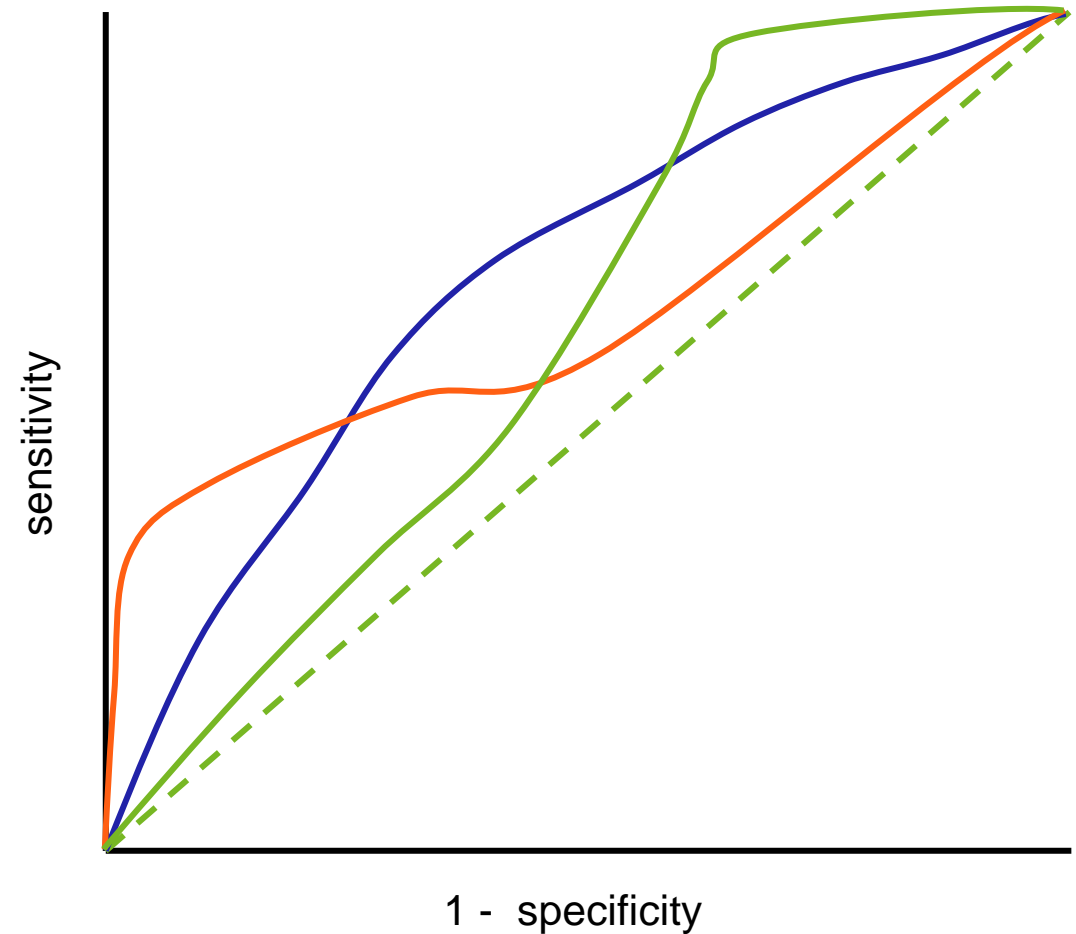
- AUC: 0.71

Biomarker B

- AUC: 0.65
- Highest specificity

Biomarker C

- AUC: 0.64
- Highest sensitivity



	Does not have the disease	Has the disease
Test is negative	True negative Cool	False negative BAD
Test is positive	False positive Okay	True positive Cool

In this case, we want **high sensitivity**



Making a biomarker for classifying abnormalities as cancer or not

Biomarker A

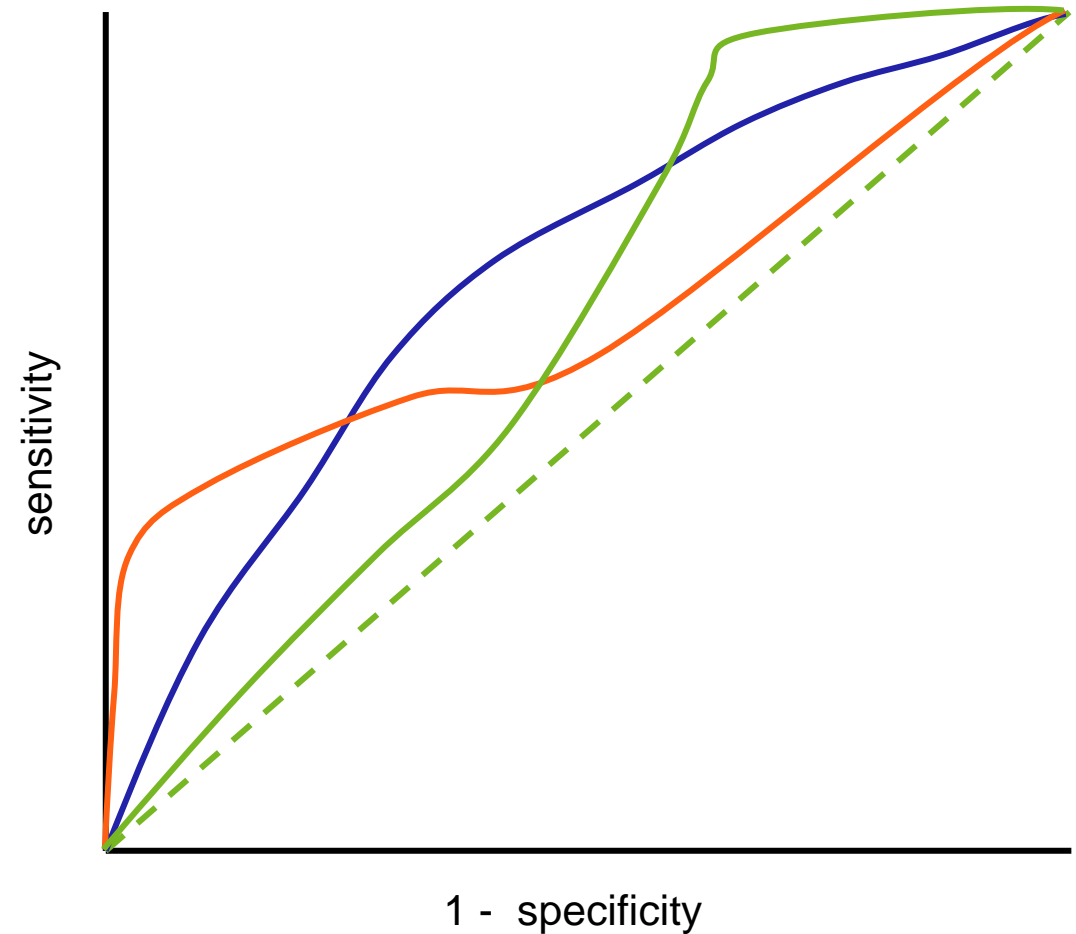
- AUC: 0.71

Biomarker B

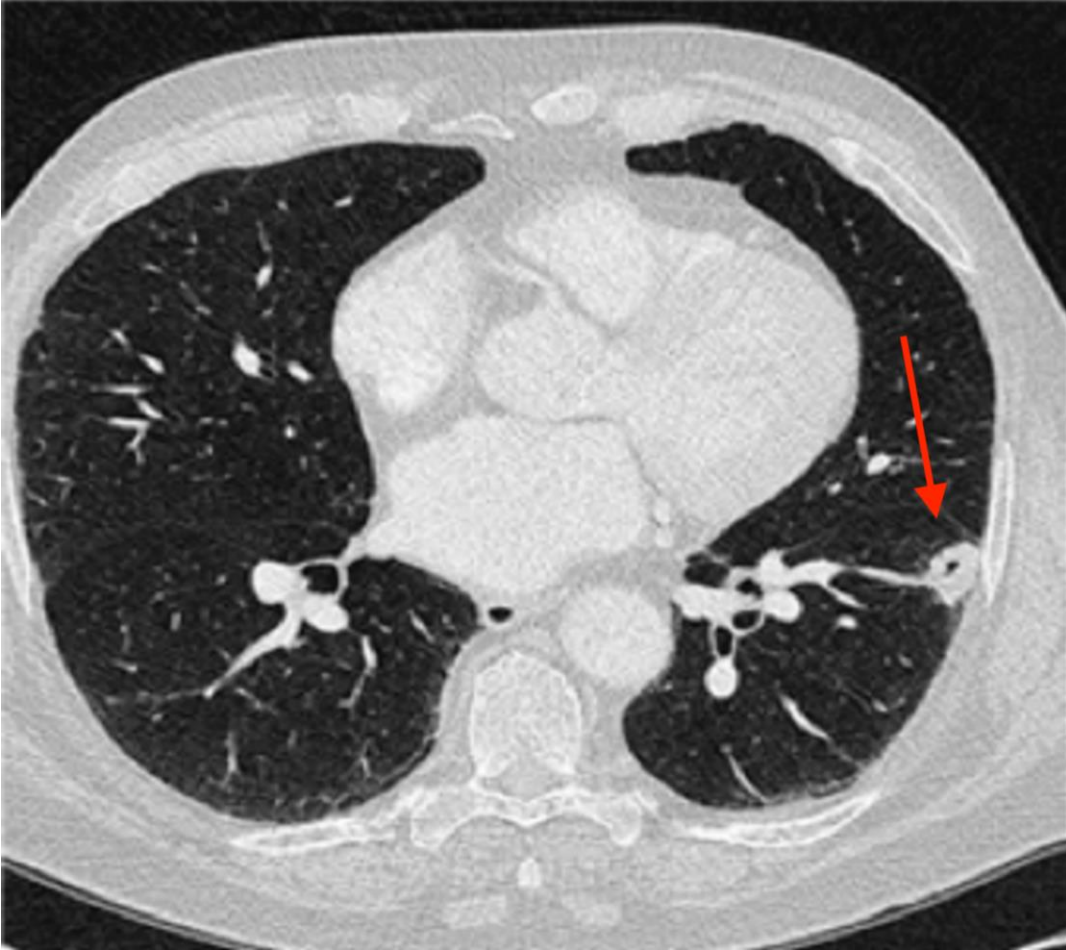
- AUC: 0.65
- Highest specificity

Biomarker C

- AUC: 0.64
- Highest sensitivity



Chest-CT shows no pulmonary embolism, but something that might be cancer



Diagnosis

- Biopsy

Treatment

- Lung surgery
- Chemotherapy

Making a biomarker for classifying abnormalities as cancer or not

Biomarker A

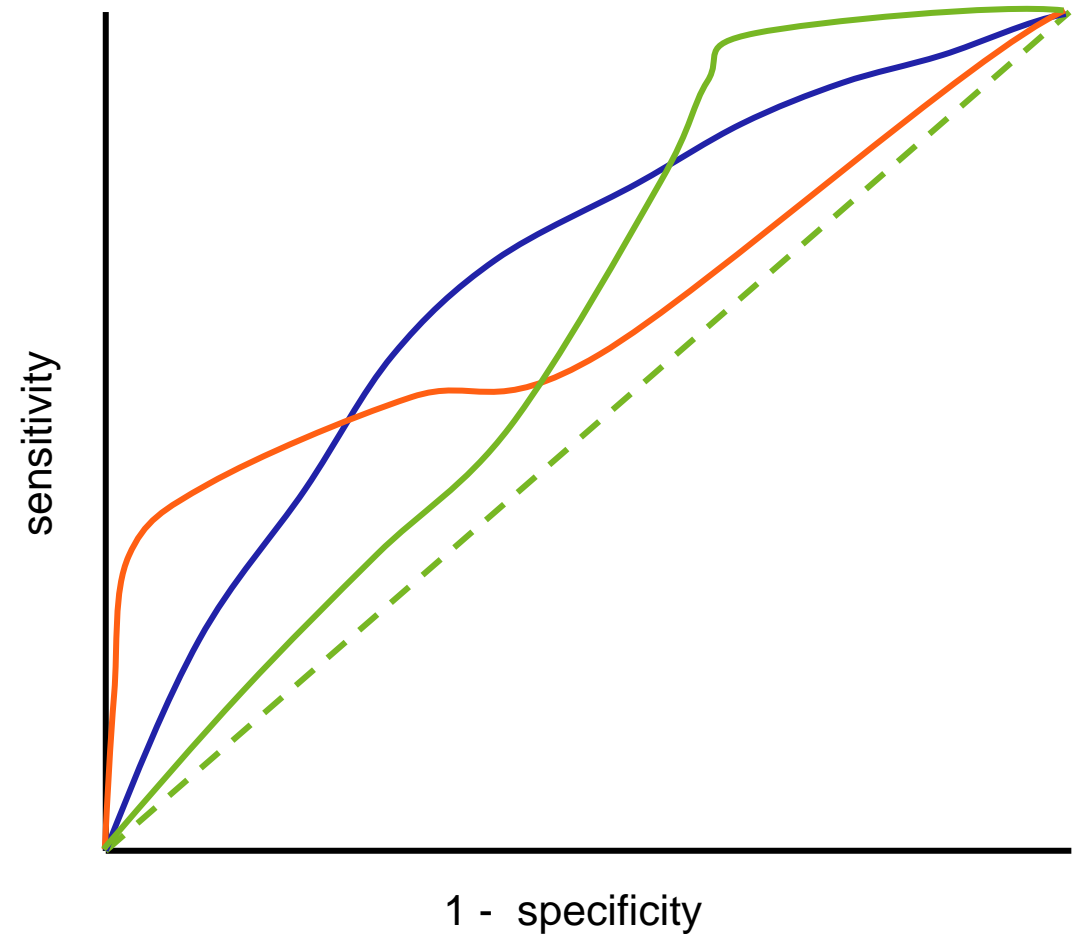
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Making a biomarker for classifying abnormalities as cancer or not

Biomarker A

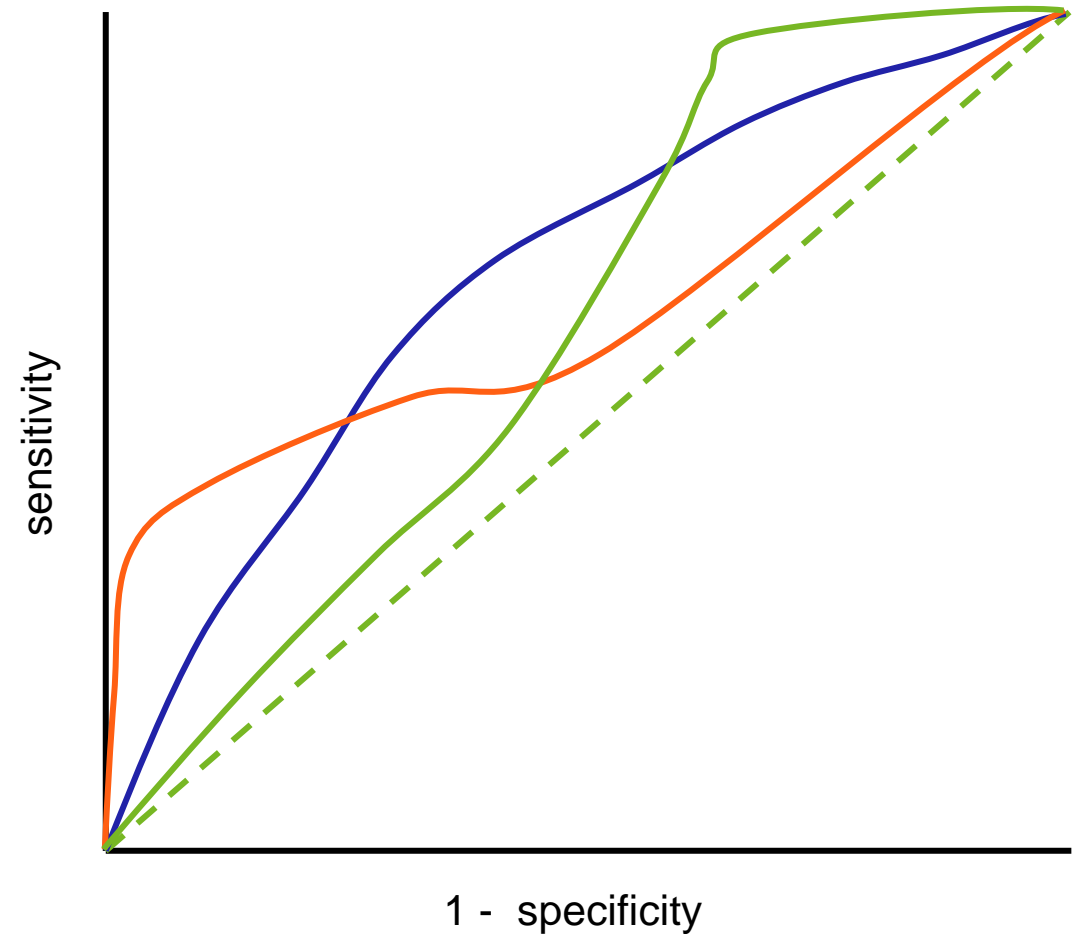
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Biomarker B

- AUC: 0.65
- Highest specificity

Biomarker C

- AUC: 0.64
- Highest sensitivity



If a **false negative** is bad, **high sensitivity** is needed

If a **false positive** is bad, **high specificity** is needed

Pathology report and other test results comes back...

Diagnosis

- Cancer, but more advanced than initially thought
- Only an experimental drug is a treatment option

Prognosis without treatment

- Median survival of 6 months

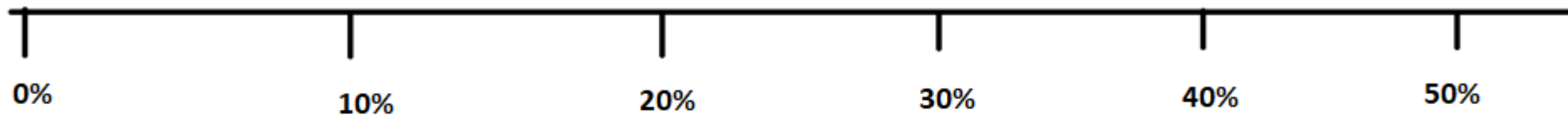
Prognosis with treatment

- 40% of patients respond and live several years
- 60% of patients have no benefit

Side-effects

- Fatigue
- Diarrhoea and/or vomiting
- Joint pain





Probability threshold (p)

More worried
about disease
progression



More worried
about side-effects



Making a biomarker for predicting response

Biomarker A

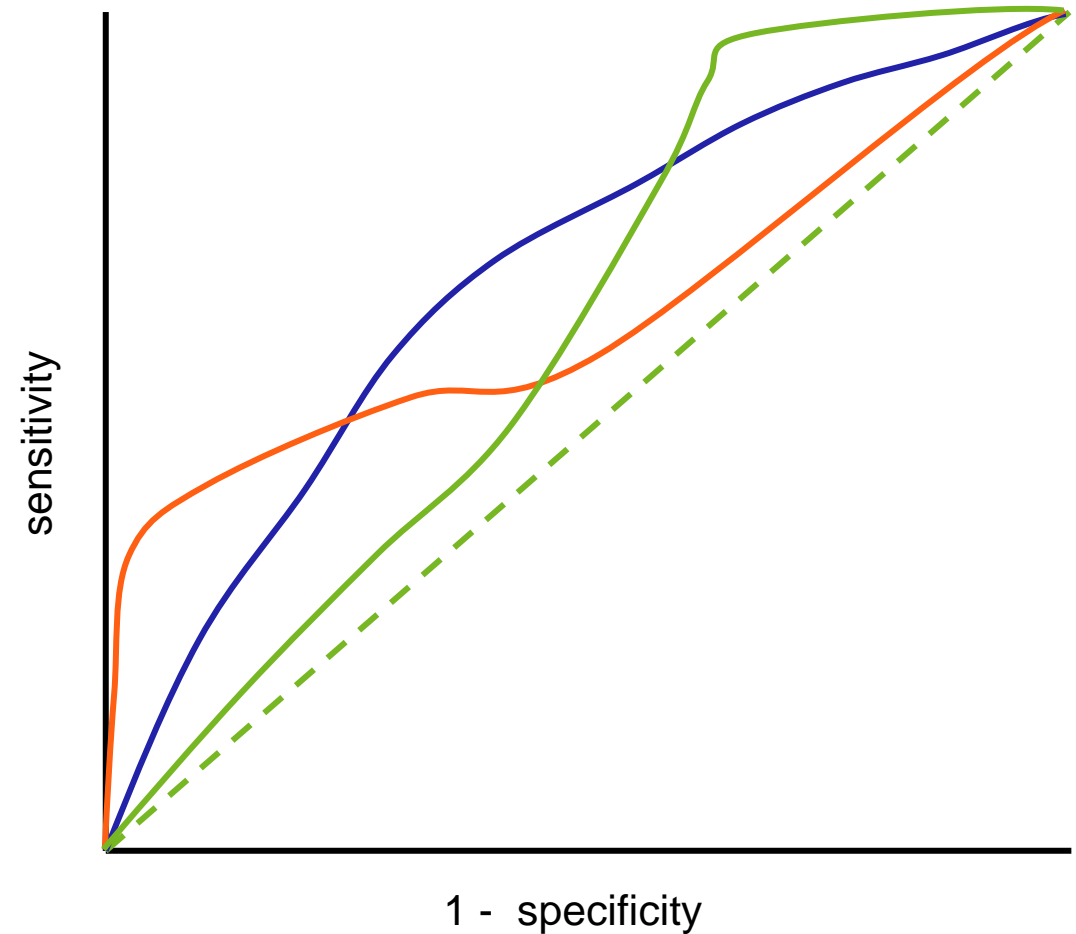
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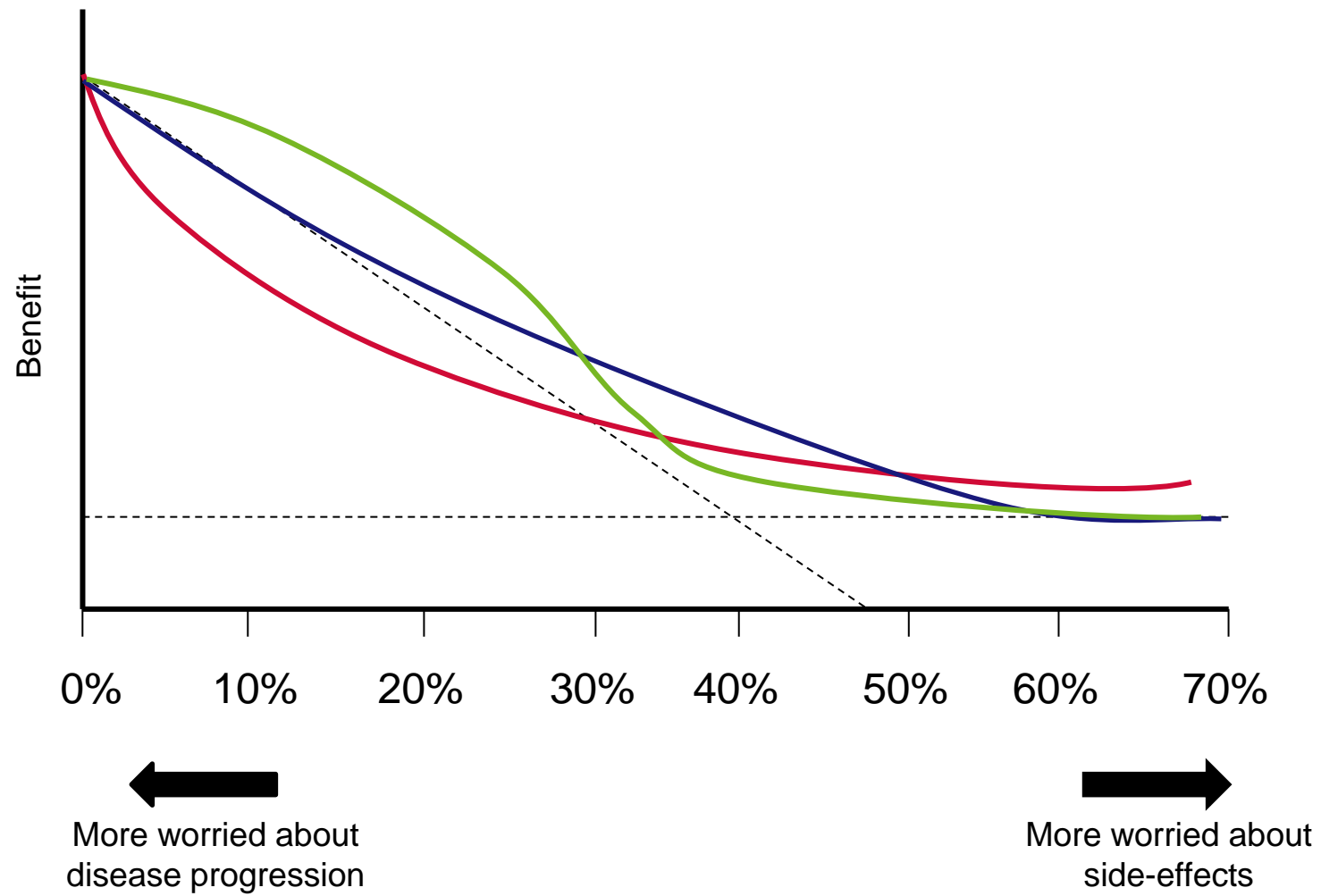
Biomarker C

- AUC: 0.64
- Highest sensitivity



How useful a biomarker is depends greatly on the preference of the patient, and these may vary considerably.





Summary

1. Explain how your model will **change decisions**.
2. **False negatives** are bad stuff? Optimize for **sensitivity**.
3. **False positives** are bad stuff? Optimize for **specificity**.
4. If patient **preferences differ** significantly, use a **decision curve**.

