

- 1. Morphology
- 2. Change (biology)
- 3. Integration of morphology & biology (genesis)
- 4. Interpretation & advancements



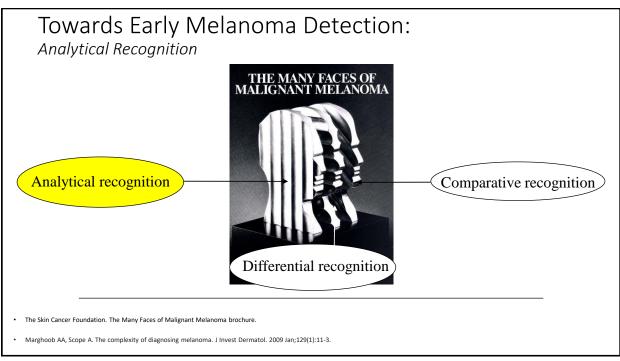


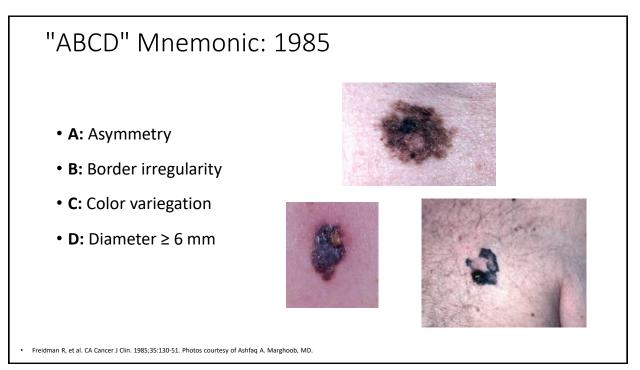
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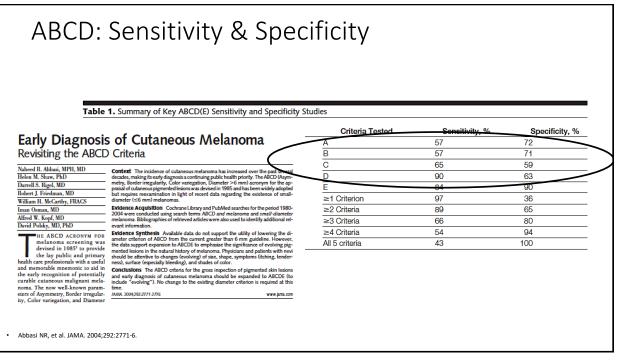
The role of the clinician in the age of technology

Morphology

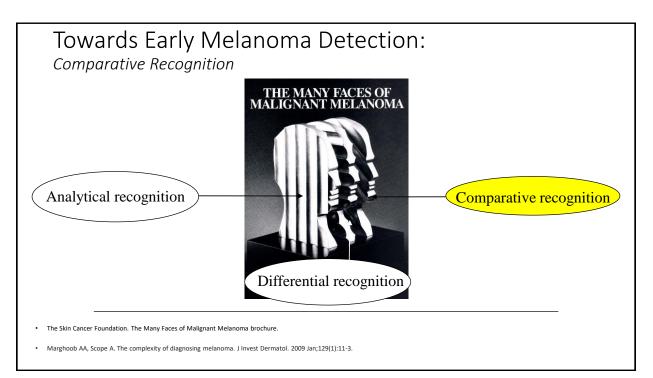
- a. Clinical expertise required to ID lesion/s at high probability for being MM
 - Humans vision: analytical, differential, & comparative recognition
 - Technology: Al Neural networks can classify lesions into risk categories
- b. Clinical expertise required to decide which lesions need further evaluation & in differentiating nevi from MM
 - ABCDs have helped but MMs can lack the clinical ABCDs or can resemble SKs
 - Dermoscopy has helped in improving DA but its not 100%
 - AI Neural networks can match expert DA
- c. Clinicians recognize that MMs can be clinically and dermoscopically featureless
 - Monitoring (biology) has helped improve DA





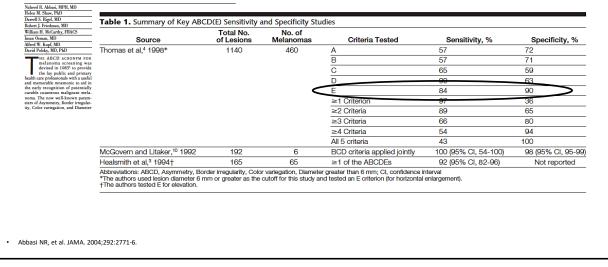




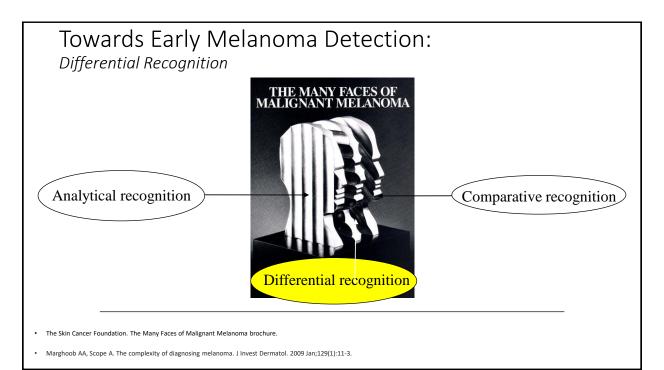


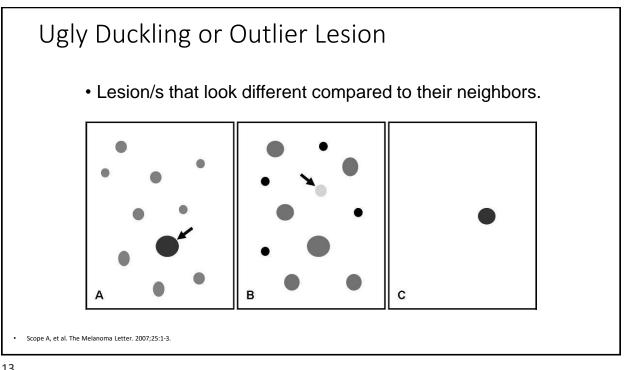
Evolution / Change: Sensitivity & Specificity

Early Diagnosis of Cutaneous Melanoma Revisiting the ABCD Criteria

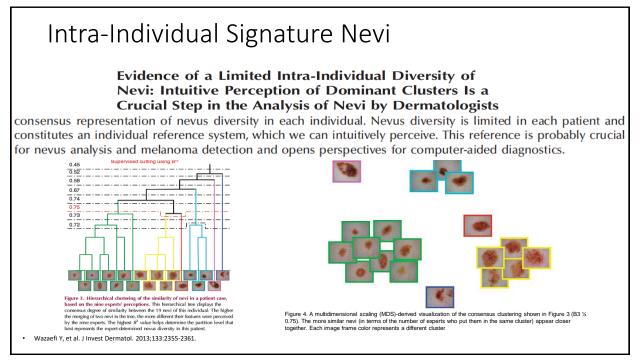


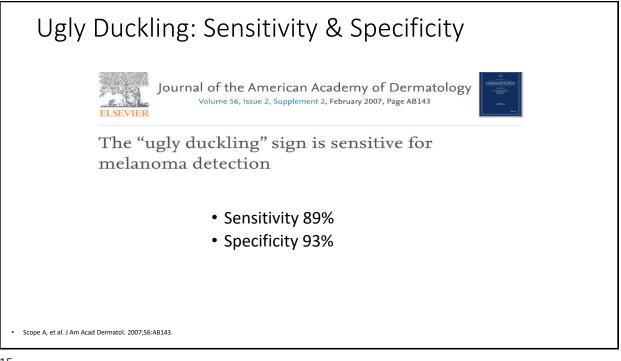












Using these basic human functions we can identify subtle MM

• Case

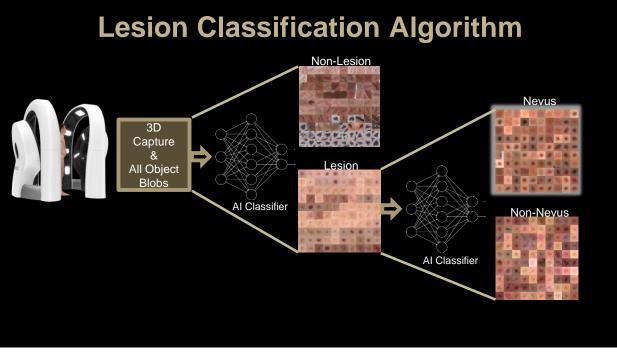
Morphology

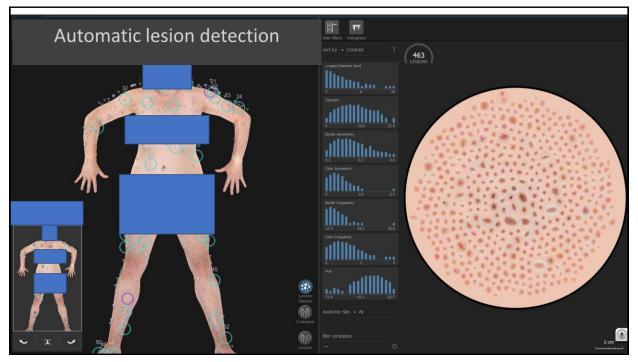
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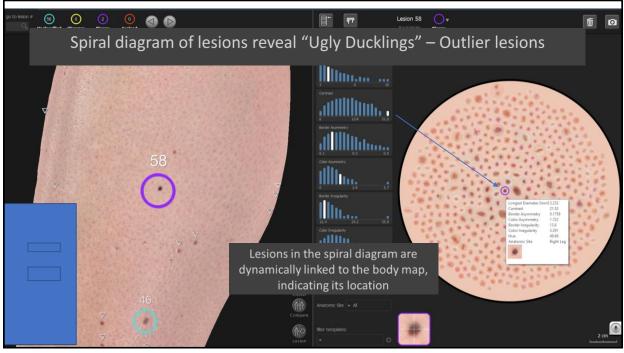


Think of technology as a supplement to your examination, not as a replacement!

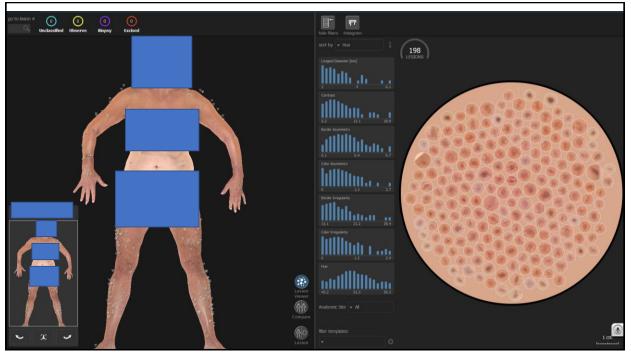
Expert & technology is not a race to see who is the best but rather a race to help forge the most synergistic partnership aimed to help improve the care we deliver to our patients!





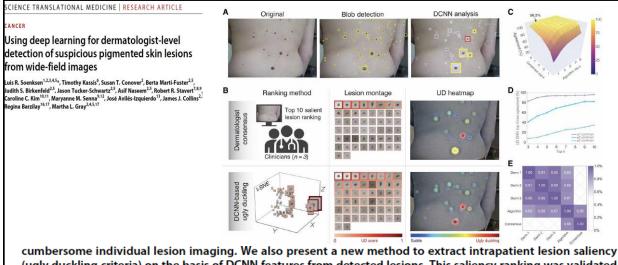




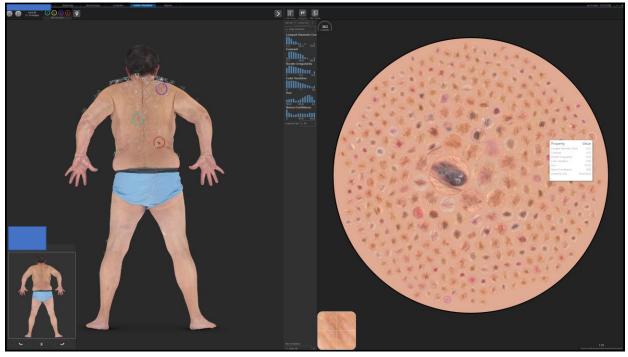


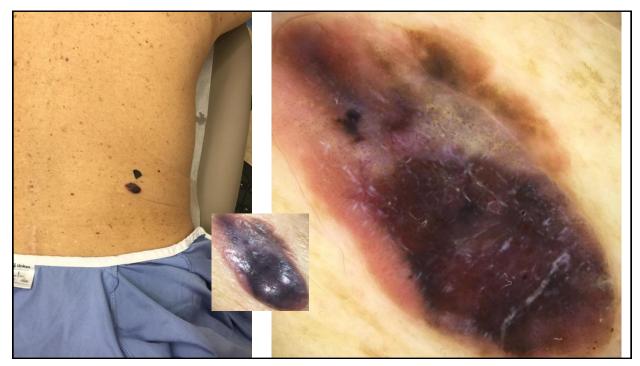






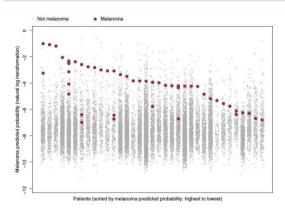
cumbersome individual lesion imaging. We also present a new method to extract intrapatient lesion saliency (ugly duckling criteria) on the basis of DCNN features from detected lesions. This saliency ranking was validated against three board-certified dermatologists using a set of 135 individual wide-field images from 68 dermatological patients not included in the DCNN training set, exhibiting 82.96% (67.88 to 88.26%) agreement with at least one of the top three lesions in the dermatological consensus ranking. This method could allow for rapid and accurate assessments of pigmented lesion suspiciousness within a primary care visit and could enable improved patient triaging, utilization of resources, and earlier treatment of melanoma.





3D Whole-body skin imaging for automated melanoma detection

Results: A total of 35 patients contributed 23,538 automatically identified skin lesions >2 mm in largest diameter (102–3021 lesions per participant). All were White patients and 23 (66%) were males. The median (range) age was 64 years (26–89). There were 49 lesions of melanoma and 22,489 lesions that were not melanoma. The AUC for the prediction model was 0.94 (95% CI: 0.92–0.96). Considering all lesions in a patient-level analysis, 14 (28%) melanoma lesions had the highest predicted score or were in the 99th percentile among all lesions for an individual patient.



2-2598-FILeOI3.jpc

Using a model-based threshold associated with 95% sensitivity for melanoma detection, the model could reduce the number of lesions requiring examination by 75%.

This means that almost 6000 lesions were classified within the MM range!!! (B:M of 1:123)

Morphology

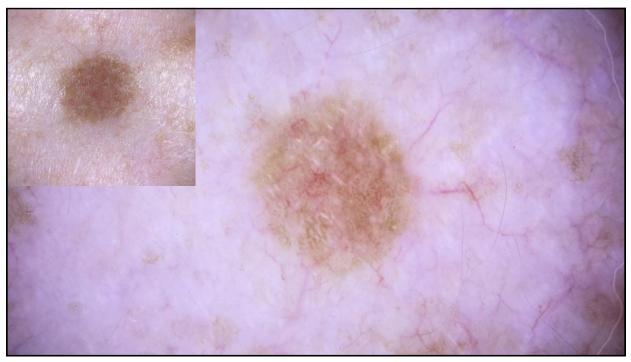
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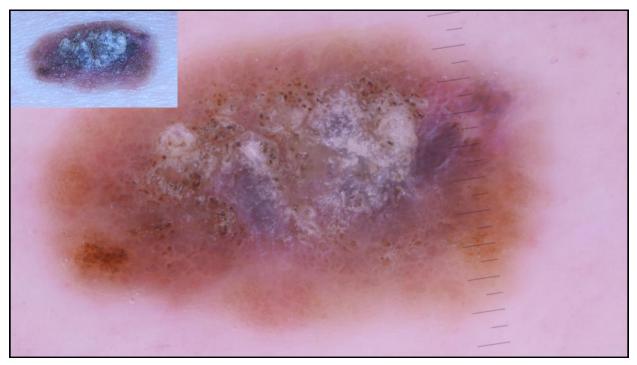


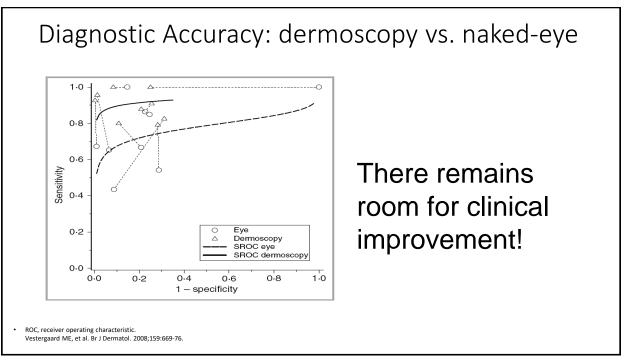


Morphology

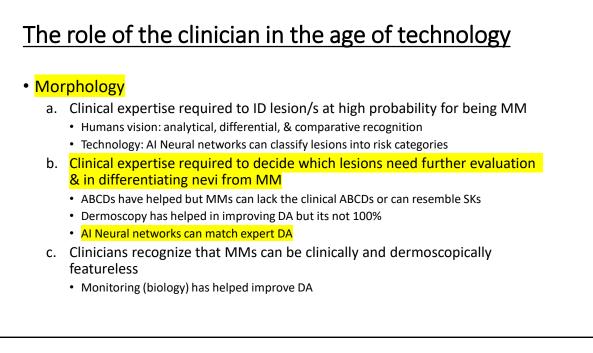
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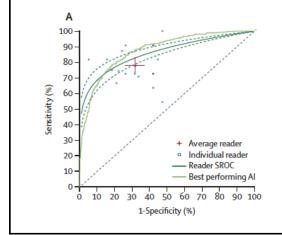






Validation of artificial intelligence prediction models for skin cancer diagnosis using dermoscopy images: the 2019 International Skin Imaging Collaboration Grand Challenge

Marc Combalia[®], Noel Codella[®], Veronica Rotemberg[®], Cristina Carrera, Stephen Dusza, David Gutman, Brian Helba, Harald Kittler, Nicholas R Kurtansky, Konstantinos Liopyris, Michael A Marchetti, Sebastian Podlipnik, Susana Puig, Christoph Rinner, Philipp Tschandl, Jochen Weber, Allan Halpern[®], Josep Malvehy[®]



Artificial intelligence for melanoma diagnosis

Philipp TSCHANDL *

teaching.⁴⁶ In conclusion, convolutional neural networks today are not inferior to dermatologists in basic image evaluation, but as rating an image is not matching a full clinical patient, or histologic, exam human-computer collaboration methodologies can be estimated to be the more successful path in the close future. For any application, the medical and scientific community should demand prospective clinical trials proving the benefit of CNNs in clinical practice.

Sensitivity similar

39

What is the significance of the insights we have made so far?

- Clinical expertise remains important and will remain so for the foreseeable future
- Technology offers the potential to elevate the expertise level of clinicians
 - Experts have a high threshold for biopsy: high specificity
 - Non-experts have a low threshold to biopsy: low specify

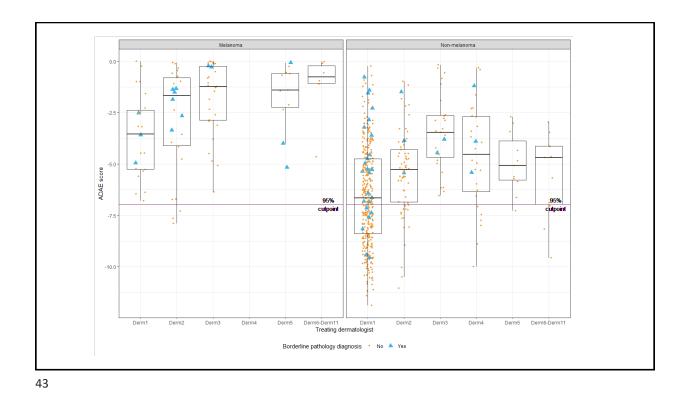
Using these basic human functions we can identify subtle MM

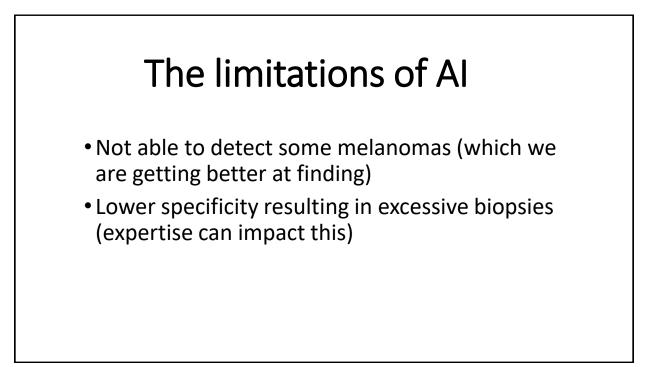
Case

41

What is the significance of the insights we have made so far?

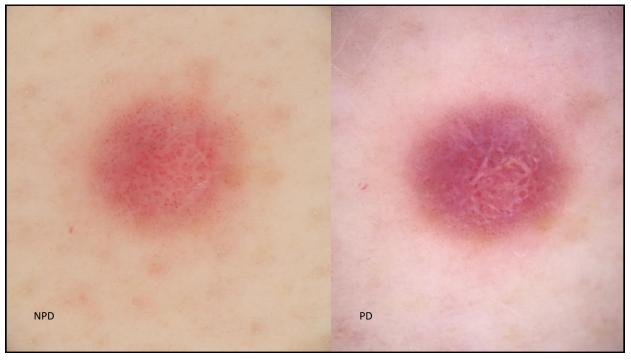
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A 50+ yo patient with skin type II & a family history of melanoma presents for skin cancer surveillance exam. This asymptomatic lesion was noted to be new.





+ - • •	Sensitivity:	0.95 • Dataset: MSK only testset • W/folds
	Nevi	
	Melanoma	
	Uploaded	< more likely nevus classification more likely melanoma >
	Map type:	heat-map 👻 Expression: regular 👻
	1.19	

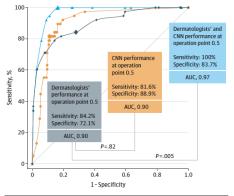
JAMA Dermatology | Original Investigation

Assessment of Diagnostic Performance of Dermatologists Cooperating With a Convolutional Neural Network in a Prospective Clinical Study Human With Machine

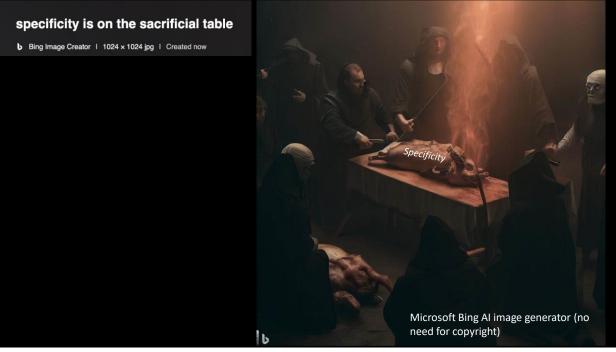
Julia K. Winker, MD: Andreas Blum, MD: Katharina Komm Albert Rosenberger, MS: Holger A. Haenssle, MD that dermatologists may improve their performance when they cooperate with the market-approved CNN and that a broader application of this human with machine approach could be beneficial for dermatologists and patients.

Table 2. Sensitivity, Specificity, and Accuracy of Diagnostic Classifications and Management Decisions of Dermatologists, Convolutional Neural Network (CNN), and Dermatologists Cooperating With CNN^a

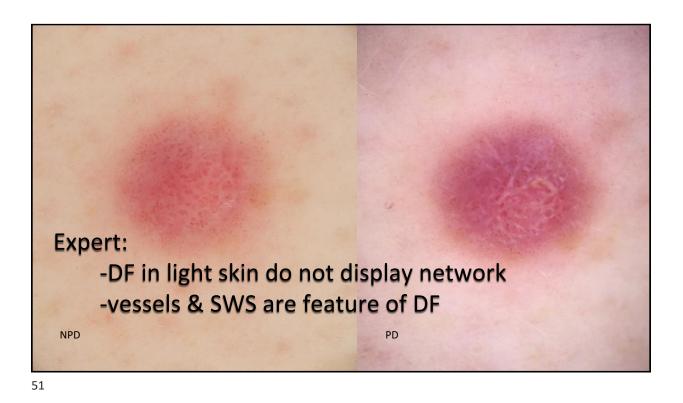
	Sensitivity	Specificity	Accuracy
Diagnostic classifications			
Dermatologists alone	84.2 (69.9-92.6)	72.1 (65.3-78.0)	74.1 (68.1-79.4)
CNN	81.6 (66.6-90.8)	88.9 (83.7-92.7)	87.7 (82.8-91.4)
Dermatologists with CNN	100.0 (90.8-100.0)	83.7 (77.8-88.3)	86.4 (81.3-90.3)
Management decisions			
Dermatologists alone	97.4 (86.5-99.5)	45.3 (38.3-52.4)	53.9 (47.5-60.3)
Dermatologists with CNN	100.0 (90.8-100.0)	55.8 (48.7-62.7)	63.2 (56.7-69.2)

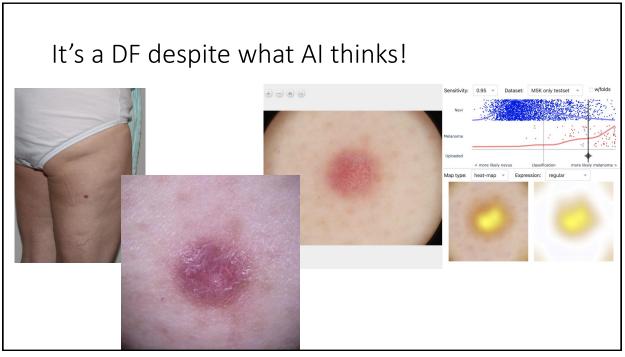


The CNN used was Moleanalyzer Pro (FotoFinder Systems). Sensitivities and specificities at the a priori operation point (cutoff for malignancy, $\geq \! 0.5)$ are depicted as larger symbols on the corresponding ROC curves. Abbreviation: AUC, area under the curve.

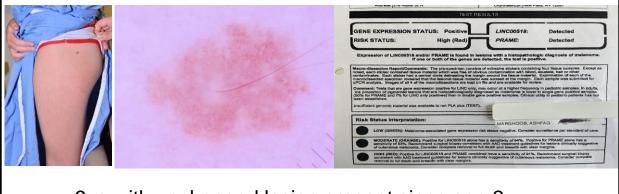








All technology from RCM, electrical impedance, Al to GEP needs to be applied to a narrow group of appropriately selected lesions!



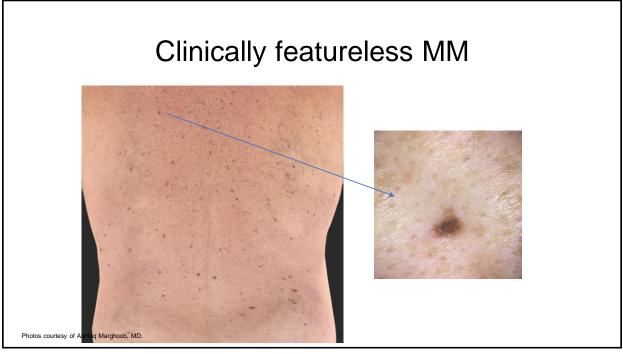
9yo with unchanged lesion present since age 2.

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The role of the clinician in the age of technology

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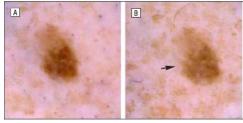
Sequential dermoscopy imaging helps in detection of socalled "featureless" melanomas

Identification of Clinically Featureless Incipient Melanoma Using Sequential Dermoscopy Imaging

Harald Kittler, MD; Pascale Guitera, MD; Elisabeth Riedl, MD; Michelle Avramidis, MD; Ligia Teban, MD; Manfred Fiebiger, MD; Rickard A. Weger, MD; Markus Dawid, MD; Scott Menzies, MBBS, PhD

Table 2. Positive Features of Melanoma per Lesion Seen in Baseline and Follow-up Images of Melanomas and Melanocytic Nevi*

No. of Positive Features of Melanoma	Melanoma	Melanocytic Nevi	P Value†
Baseline images	n = 91	n = 408	
0	56 (61.5)	279 (68.4)	.41
1	28 (30.8)	107 (26.2)	
>1	7 (7.7)	22 (5.4)	
Follow-up images, 1.5-4.5 mo	n = 34	n = 202	
0	21 (61.8)	152 (75.2)	.15
1	11 (32.4)	41 (20.3)	
>1	2 (5.9)	9 (4.5)	

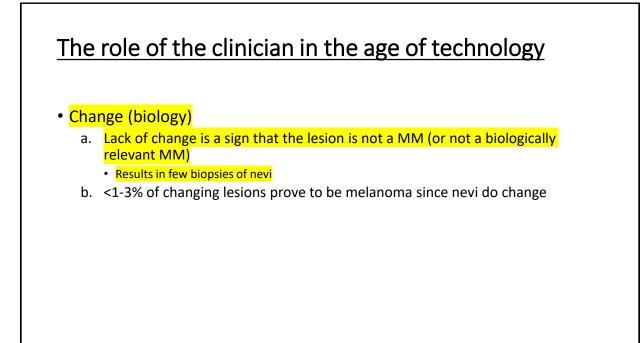


- 61.5% of melanomas had no MMspecific structures at baseline.
- 61.8% of melanomas had no MMspecific structures at f/u of between 1.5-4.5 months!
- Only global changes helped identify these 'insipient' melanomas

ARCH DERMATOL/VOL 142, SEP 2006 WWW.ARCHDERMATOL.COM 1113

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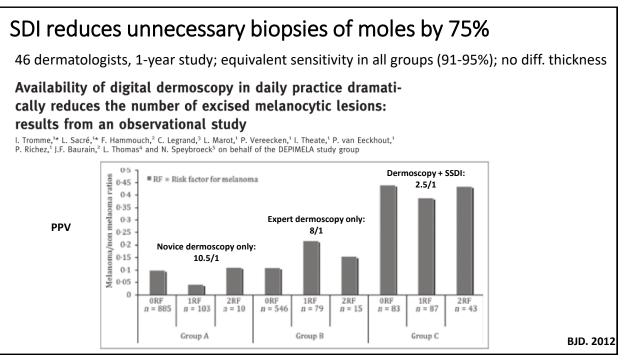


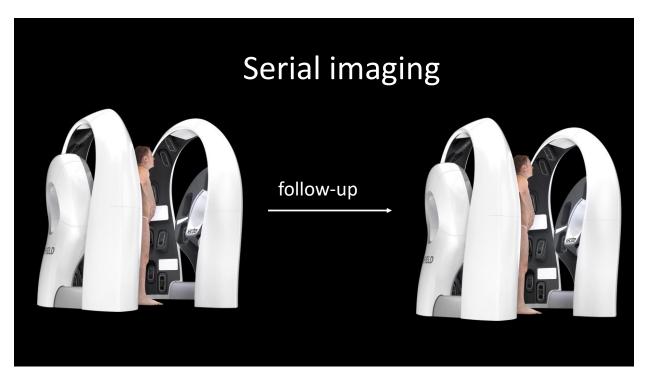


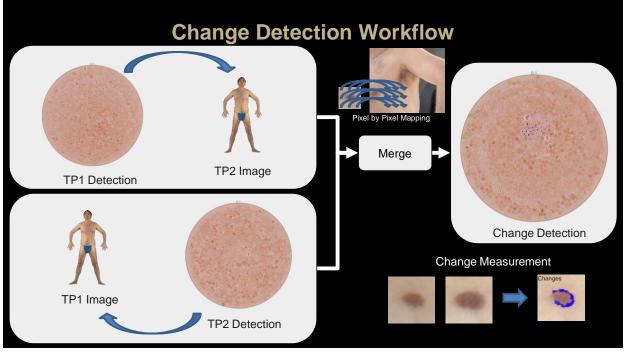


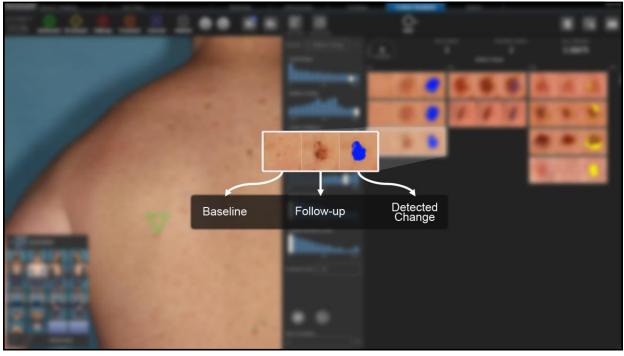
TBP: redu	ces the number o	of unnecessary b	iopsies		
Photographic Monito	s of Total Body and Dermatoscop oring of Nevi in Similar Patient for Cutaneous Melanoma	ic			
Agnessa Gadeliya Goodson, Mi Glen M. Bowen, MD,*† and Do	D,* Scott R. Florell, MD,* Mark Hyde, PA-C,† Juglas Grossman, MD, PhD*†	-Fewer nevi	biopsied per patient		
luminescence microscopy (DELM) mas presenting as new lesions or OBJECTIVE To determine whether gin (de novo vs nevus derived) dii photography. METHODS ON METHODS whethousand sevent tography and were monitored usi	erience monitoring new in high-risk patients using seri photography schieved low biopsy rates but was limits arising from new that had not been photographed. It biopsy rates, efficiency of melanoma detection, and n fered in a similar patient population monitored using to y-six patients (including 187 from a prior cohort) under ng photographs obtained at the initial visit. Risk factors lents were comparable with those of patients previous	d by melanoReduces th helanoma ori- tat body (TB) biopsied for went TB pho- s and median	biopsied for each MM found		
RESULTS Two hundred seventy- melanomas detected on follow-u new lesions, nine arose de novo,	ive biopsies were performed in 467 patients on follow-u	No TBP	ТВР		
CONCLUSIONS In our experienc ing TB photography was associat using DELM and facilitated dete	Goodson et al	+dermoscopy	+dermoscopy		
melanomas detected on follow-u	Biopsies on follow up	1.1 per pt	0.59 per pt		
	Nevus:MM	45	17		
	% MM de novo	83%	75%		
total body photography. Ison AG, et al. Dermatol Surg. 20	010;36:1087-98.				
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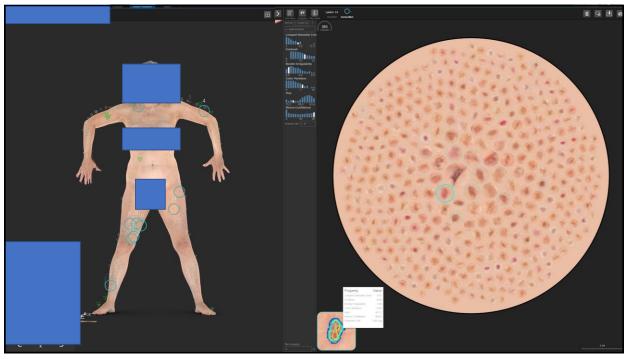


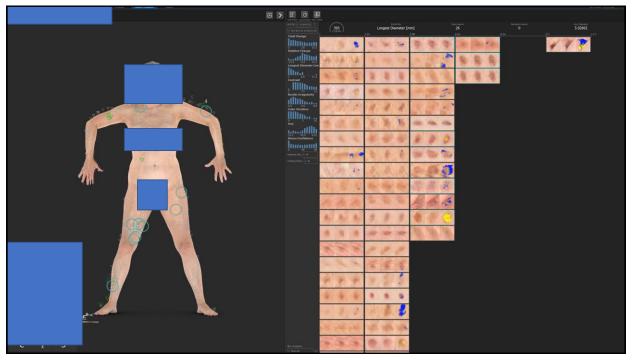


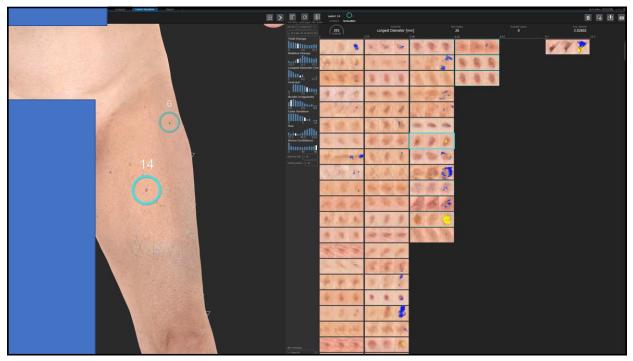




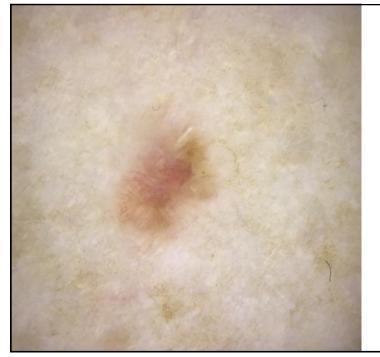












 Dermoscopy remains the main instrument to help decide which changed lesions require further investigation.

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The role of the clinician in the age of technology

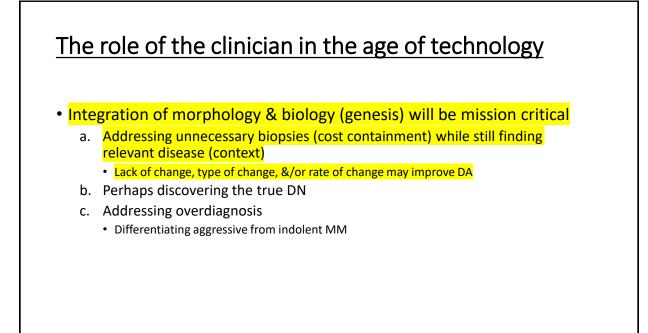
• Change (biology)

- a. However, lack of change is a sign that the lesion is not a MM (or not a biologically relevant MM)
- b. <1-3% of changing lesions prove to be melanoma since nevi do change

Study	Year	# pts.	# lesions followed	Nevi followed/ patient	# changed	# melanoma	%MM/ changed
Schiffner	2003	145	272	1.9	95	0	0.0
Bauer	2005	196	2015	10.3	128	2	1.6
Robinson	2004	100	3482	34.8	193	4	2.1
Banky	2005	309			573	18	3.1

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Type of change

- Nevi change:
 - Nevi with peripheral globules
 - Nevi with streaks
- Melanoma change
 - Type of change and not necessarily the rate of change is key

Streaks = radial growth phase of Reed's nevi



Argenziano G, Agozzino M, Bonifazi E, et al. Natural Evolution of Spitz Nevi. *Dermatology.* 2011.

Peripheral globules = radial growth phase of some moderate DN (LAN)



Frequency and Characteristics of Enlarging Common Melanocytic Nevi Hand Kuler, MD. Merkus Steinheim, MD. Matheward, MD. Habers Pehankerger, MD, Kanne WG, MD, PEPP, Moled Block, MD

Manhan To anybra the features and characteric 1. The features

stager Cabott study using adjutal epideministensive encocycy/TEMD for documentation and follow-up, with median follow-up interval of 11.4 months.
stettings A dermatology department at a university hosion in Viruna, A constaint, and a study of the study of the net appearing efficiency in a common new, obtained in 1859 patienties (mean [aSD] age, 34.2 a 14.8 y; 35.6% mb(e).

rventions Fellow up examination and documenn by digital ELM. In Outcome Measuress Prequency of cularging next rding to age and comparison of ELM features obed in enlarging and nonenlarging nevi. The frequency of enlarging next was inversely related to one of the second in only 0.75 second second second second $P_{\rm c}(M)$. This provides the second second second second in only 0.75 second enlarging next relation $P_{\rm c}(M)$. This provides the second $P_{\rm c}(M)$. This provides the second seco

devices in the trequency of enlarging common in versely related to age. In the absence of clinical sky ypia, enlargement alone does not indicate mai y. A peripheral tim of brown globules is a chan the ELM feature of symmetrically enlarging mela

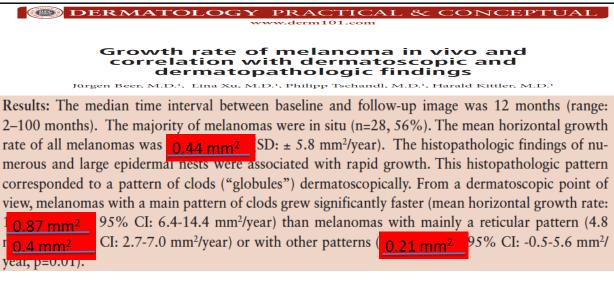
matol. 2000;136:316-320



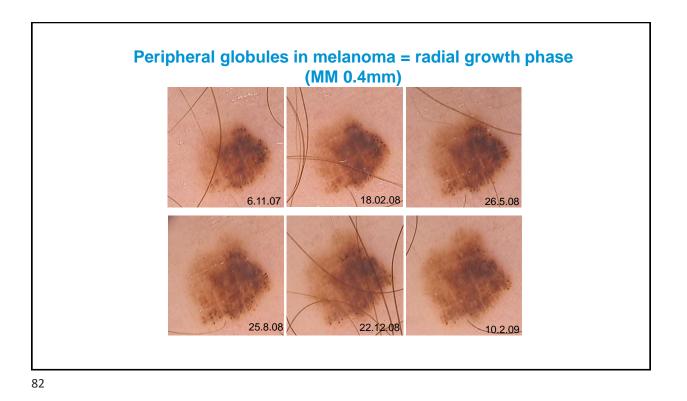
Original Investigation Growth-Curve Modeling of Nevi With a Peripheral Globular Pattern Shirin Bajaj, BA: Stephen W. Dusza, DrPH; Michael A. Marchetti, MD: Xinyuan Wu, BA: Maira Fonseca, BS; Kivanc Kose, PhD; Johanna Brito, MD; Cristina Carrera, MD: Vanessa P. Martins de Silva, MD; Josep Malvehy, MD; Susana Puig, MD, PhD; Sarah Yagerman, MD; Tracey N. Liebman, MD; Alon Scope, MD; Allan C. Halpern, MD, MSc; Ashfaq A. Marghoob, MD 100 **Growth Rate** CI, p-value MSKCC MSKCC (fitted) (per month) Barcelona SONIC Barcelona (fitted) SONIC (fitted) 80 (mmb)2) MSKCC 0.52mm² 95% CI: 0.43-0.61, p<0.001 And Barcelona 0.12mm^2 95% CI: 0.03-0.21, p=0.007 00 20 SONIC 0.13mm^2 95% CI: 0.01-0.24, p=0.033 0 72 onths) $0.27 mm^{2}$ 95% CI: 0.15-0.39, p=0.033 96 108 120 Average 60 84

In those lesions that stopped growing ($\sim 20\%$), median time to growth cessation was at least 49.5 months, thus if a clinician

80



PG nevi growth rate: average 0.12-0.52mm² (depending on cohort) with fastest growth recorded at 2.26mm² !



Horizontal rate of growth of PG nevi and melanoma is the same

Melanoma		
Grow asymmetrically (63%)		
Often develop new features		
Globules do not disappear & if they do they are replaced by another MM specific structures		
No growth cessation		
BJD British Journal of Dermatology		

Integration of multiple technologies to evaluate suspect lesions

Clinical level [a,b]

- Confocal
- OCT
- Multi-spectra (SIAscopy)

Mechanical level [b]

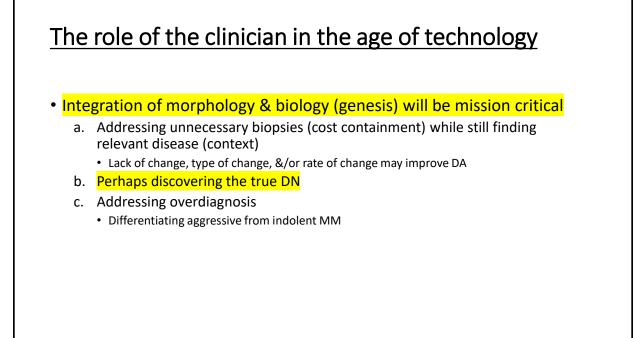
Electrical impedance

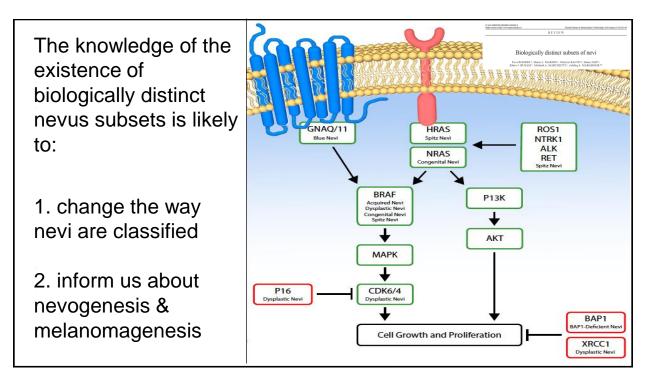
Molecular level signatures

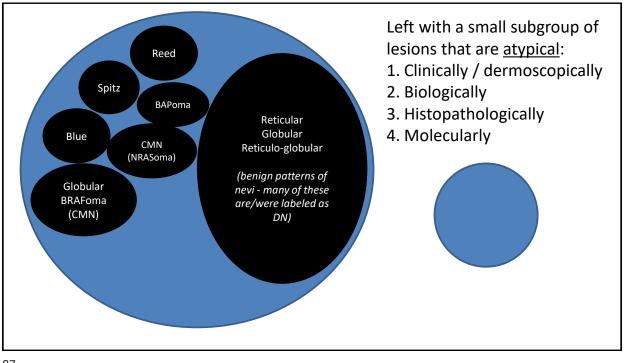
- Tape strip for RNA^[c]
- Dual photon spectroscopy (Raman)^[d]

- Example:
 - Changing lesion identified
 - Use GEP
 - If neg: follow
 - If pos: biopsy

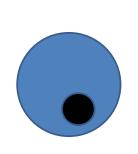
OCT, optical coherence tomography. a. Schuh S, et al. Cancers. 2022;14:1140. b. Ferrante di Ruffano, et al. Cochrane Database Syst Rev. 2018; 12:CD013186;c. Yao Z, et al. J Drugs Dermatol. 2017;16:979-986; d. Serebrennikova KV, et al. Biosensors (Basel). 2021;11:512. These materials or any continu the set of extracting deverged tree Any commercial use or distribution of these materials or any continu thereof is strictly prohibited.





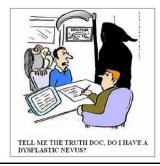


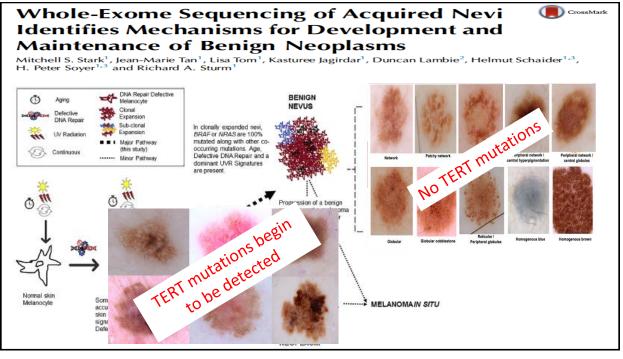




Questions :

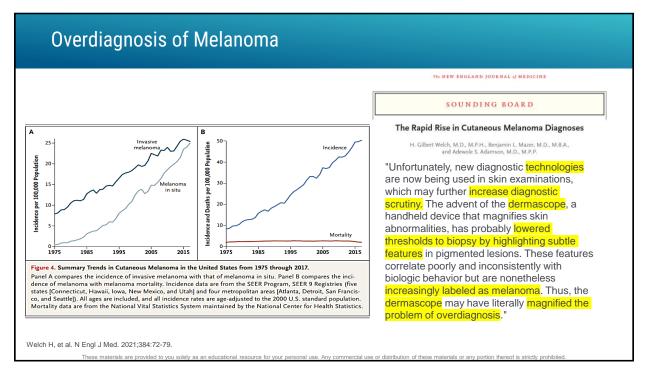
- Could a subset of these atypical lesions be the true DN (intermediary) lesion that has fueled so much debate over the decades?
- 2. Could a higher proportion of these nevi be on the evolutionary path to MM (precursor)?





<u>The role of the clinician in the age of technology</u>

- Integration of morphology & biology (genesis) will be mission critical
 - a. Addressing unnecessary biopsies (cost containment) while still finding relevant disease (context)
 - Lack of change, type of change, &/or rate of change may improve DA
 - b. Perhaps discovering the true DN
 - c. Addressing overdiagnosis
 - Differentiating aggressive from indolent MM



Estimating the magnitude of cancer overdiagnosis in Australia

Paul P Glasziou¹, Mark A Jones¹, Thanya Pathirana², Alexandra L Barratt³, Katy JL Bell³

Cancer	Lifetime probability of diagnosis		Change in probability,	Overdiagnosis	Cancer diagnoses,	Estimated overdiagnoses,
	1982*	2012	1982-2012	proportion [†]	2012	2012 (95% CI)
Prostate	0.115	0.197	0.082	42%	20 759	8636 (8281–8991)
Melanoma						
Including in situ carcinomas	0.059	0.139	0.080	58%	14 436	8315 (8034–8596)
Invasive only	0.054	0.069	0.015	22%	7151	1552 (1331–1773)
Thyroid	0.0014	0.0052	0.0038	33%	661	483 (426–540)
Renal	0.011	0.019	0.008	42%	2045	861 (749–973)
Other invasive cancers	0.420	0.423	0.003	_	39 452	_

Indolent vs Aggressive Melanomas



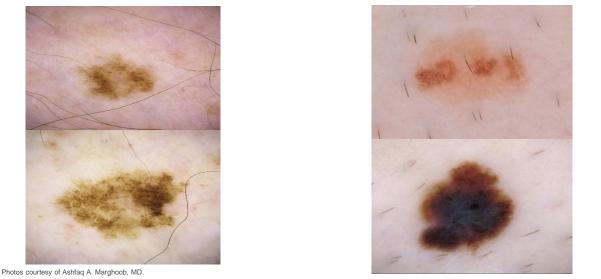
4 years before invasive MM and metastasis developed (+SLNB)

SNLB, sentinel lymph node biopsy. Photos courtesy of Ashfaq A. Marghoob, MD.

MM 0.2 mm: not NAM

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Can be predict which melanoma will become aggressive?



Rate of growth of Melanoma: Dermoscopy

DERMATOLOGY PRACTICAL & CONCEPTUAL

www.derm101.com

Results: The median time interval between baseline and follow-up image was 12 months (range: 2–100 months). The majority of melanomas were in situ (n=28, 56%). The mean horizontal growth rate of all melanomas was 5.3 mm²/year (SD: \pm 5.8 mm²/year). The histopathologic findings of numerous and large epidermal nests were associated with rapid growth. This histopathologic pattern corresponded to a pattern of clods ("globules") dermatoscopically. From a dermatoscopic point of view, melanomas with a main pattern of clods grew significantly faster (mean horizontal growth rate: 10.4 mm²/year, 95% CI: 6.4-14.4 mm²/year) than melanomas with mainly a reticular pattern (4.8 mm²/year, 95% CI: 2.7-7.0 mm²/year) or with other patterns (2.6 mm²/year, 95% CI: -0.5-5.6 mm²/year, 9=0.01).

Beer J, et al. Dermatol Pract Concept. 2011;1:59-67.

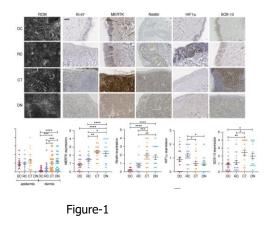
96

Predicting Aggressiveness: RCM criteria

In Vivo Melanoma Cell Morphology Reflects Molecular Signature and Tumor Aggressiveness

Alessandra Marconi^{1,7}, Marika Quadri^{1,7}, Francesca Farnetani², Silvana Ciardo², Elisabetta Palazzo¹, Roberta Lotti¹, Anna Maria Cesinaro³, Luca Fabbiani⁴, Cristina Vaschieri¹, Mario Puviani⁵, Cristina Magnoni², Shaniko Kaleci², Carlo Pincelli¹ and Giovanni Pellacani^{2,6}

Melanoma is the deadliest type of skin cancer characterized by high cellular heterogeneity, which contributes to therapy resistance and unpredictable disease outcome. Recently, by correlating reflectance confocal microscopy morphology with histopathological type, we identified four distinct melanoma subtypes: dendritic cell, round cell, demal nest, and combined-type melanomas. In this study, each reflectance confocal microscopy melanoma subtype expressed a specific biomolecular profile and biological behavior in vitro. Markers of tumor aggressiveness, including Ki-67, MERTk, nestin, and stemness markers were highest in the most invasive combined-type and dermal nest melanomas than in dendritic cell and round cell melanomas. This was also confirmed in multicellular tumor spheroids. Transcriptomic analysis showed modulation of cancer progression-associated genes from dendritic cell to combined-type melanomas. This in reconstructs. It displayed a unique behavior and a molecular profile associated with a high degree of aggressiveness. Altogether, our results show that each reflectance confocal microscopy melanoma subtype has a distinct biological and gene expression profile related to tumor aggressiveness, confirming that reflectance confocal microscopy can be a dependable tool for in vivo detection of different types of melanoma and for early diagnostic screening. *Jound of messigner benoming* 202111;2024



Marconi A, et al. J Invest Dermatol. 2022;142:2205-2216.

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Identifying the fast-growing melanomas: Molecular level

Molecular characterization of fast-growing melanomas

Caroline Gaudy-Marqueste, MD, PhD,^a Nicolas Macagno, MD, PhD,^b Anderson Loundou, PhD,^c Eric Pellegrino, MSc,^d L'houcine Ouafik, PhD,^d Timothy Budden, PhD,^e Piyushkumar Mundra, PhD,^f Gabriela Gremel, PhD,^f Victoria Akhras, MD,^g Lijing Lin, PhD,^h Martin Cook, MD,^f Rajiv Kumar, PhD,ⁱ Jean-Jacques Grob, MD,^a Eduardo Nagore, MD, PhD,^j Richard Marais, PhD,^f and Amaya Virós, MD, PhD^e Marseille, France; Manchester and London, United Kingdom; Heidelberg, Germany; València, Spain

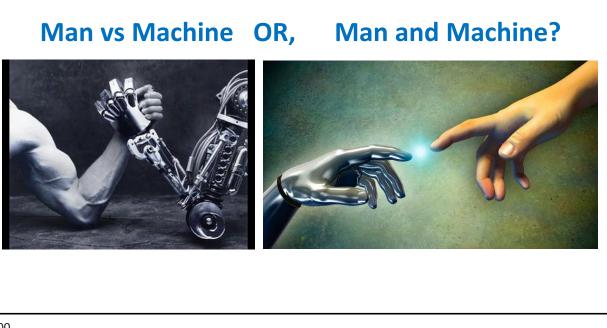
Results: Two hundred patients were enrolled, among wom 70 had FGMM. The relapse-free survival was lower in the FGMM group (P = .014). FGMM had a higher number of predicted deleterious mutations within the 40 genes than nonFGMM (P = .033). Ulceration (P = .032), thickness (P = .006), lower sun exposure (P = .049), and fibroblast growth factor receptor 2 (*FGFR2*) mutations (P = .037) were significantly associated with fast growth.

Gaudy-Marqueste C, et al. J Am Acad Dermatol. 2022;86:312-321

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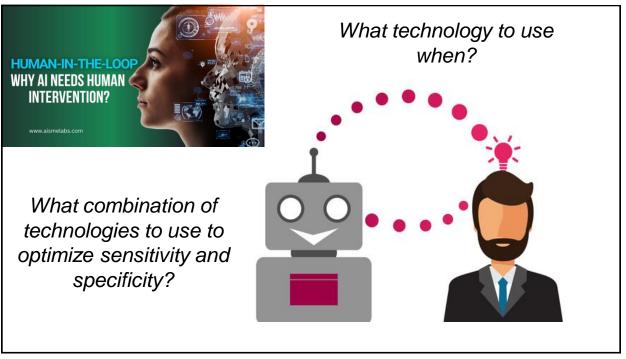
The role of the clinician in the age of technology

- 1. Morphology
- 2. Change (biology)
- 3. Integration of morphology & biology (genesis)
- 4. Interpreter & advancements





Technology can help <u>risk stratify patients</u>, <u>risk stratify lesions</u> into low risk and high/unknown risk categories OR <u>direct human vision</u> to areas to pay attention to!

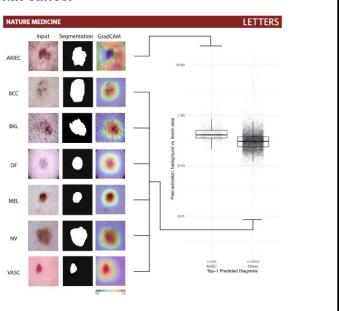


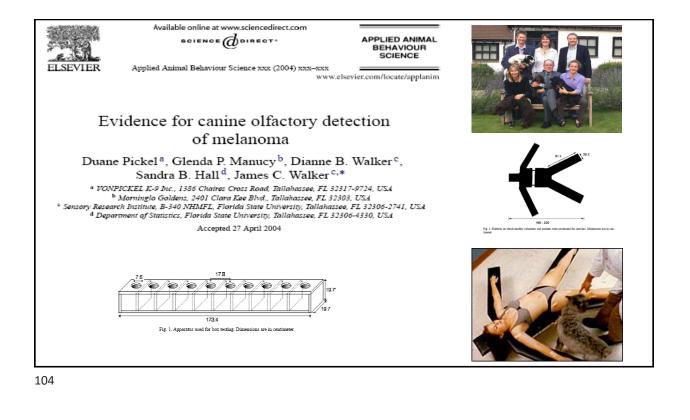
Human-computer collaboration for skin cancer recognition

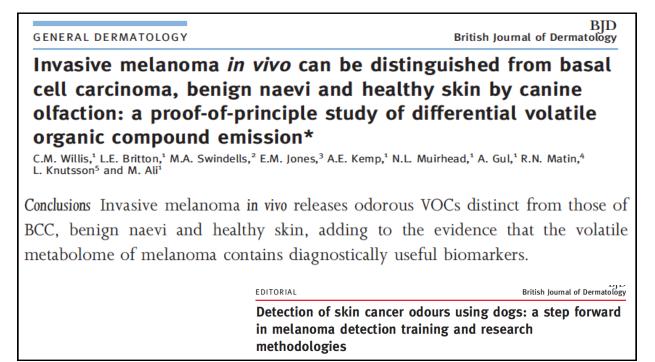
Philipp Tschandl¹⁷, Christoph Rinner^{2,37}, Zoe Apalla³, Giuseppe Argen Allan Halpern⁶, Monika Janda⁷, Aimilios Lallas³, Caterina Longo^{8,9}, Josep I Susana Puig^{10,11}, Cliff Rosendahl¹⁴, H. Peter Soyer¹⁵, Iris Zalaudek¹⁶ and H

Al was consistently better at diagnosing AK compared to physicians.

The AI based its decision on background actinic damaged skin.









"Discovery consists of seeing what everyone else has seen and thinking what no one else has thought."

Albert Szent-Giorgyi von Nagyrapolt Nobel Prize laureate, 1937



