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## REVIEW ARTICLE

# Measurement of breast density with digital breast tomosynthesis—a systematic review

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## ABSTRACT

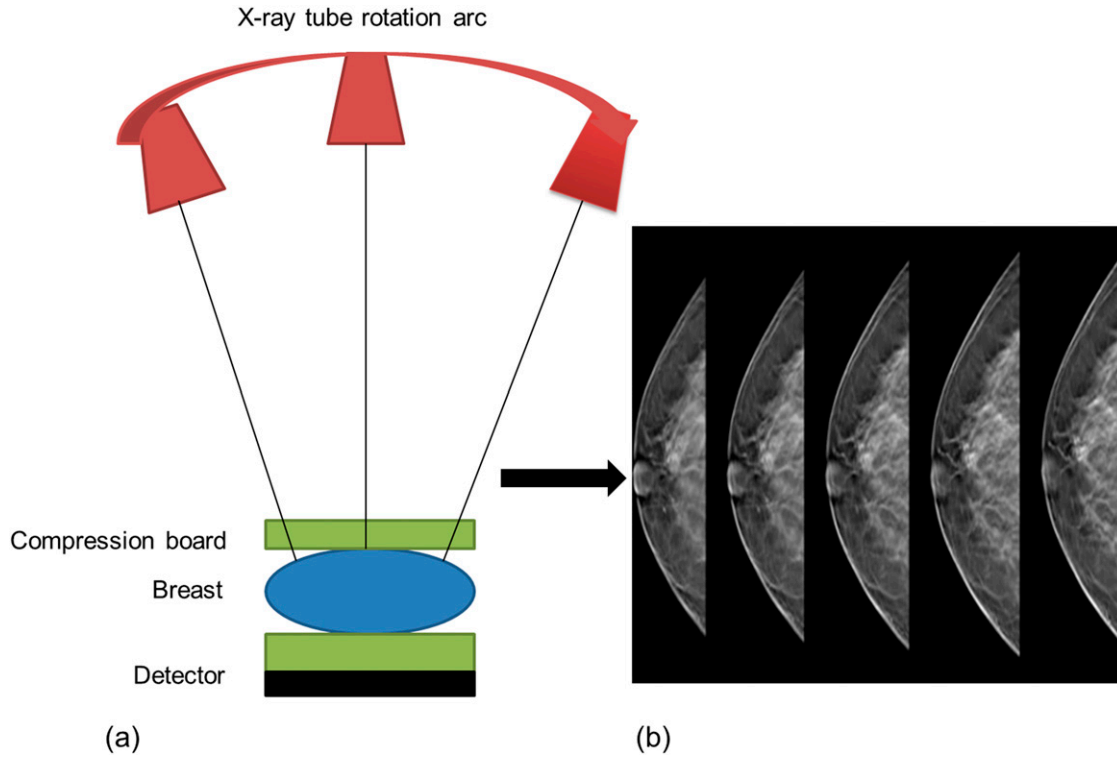
Digital breast tomosynthesis (DBT) has gained acceptance as an adjunct to digital mammography in screening. Now that breast density reporting is mandated in several states in the USA, it is increasingly important that the methods of breast density measurement be robust, reliable and consistent. Breast density assessment with DBT needs some consideration since quantitative methods are modelled for two-dimensional (2D) mammography. A review of methods used for breast density assessment with DBT was performed. Existing evidence shows Cumulus has better reproducibility than that of the breast imaging reporting and data system (BI-RADS®) but still suffers from subjective variability; MedDensity is limited by image noise, whilst Volpara and Quantra are robust and consistent. The reported BI-RADS inter-reader breast density agreement ( $k$ ) ranged from 0.65 to 0.91, with inter-reader correlation ( $r$ ) ranging from 0.70 to 0.93. The correlation ( $r$ ) between BI-RADS and Cumulus ranged from 0.54–0.94, whilst that of BI-RADS and MedDensity ranged from 0.48–0.78. The reported agreement ( $k$ ) between BI-RADS and Volpara is 0.953. Breast density correlation between DBT and 2D mammography ranged from 0.73 to 0.97, with agreement ( $k$ ) ranging from 0.56 to 0.96. To avoid variability and provide more reliable breast density information for clinicians, automated volumetric methods are preferred.

Breast cancer accounts for approximately 23% of all cancers in females and is the most frequent cause of cancer deaths in females worldwide.<sup>1–3</sup> The exact aetiology of the disease is complex, but many risk factors have been documented in the literature amongst which is breast density.<sup>4–7</sup> Breast density refers to the proportion of the breast that is composed of fibroglandular tissue. Breasts with high density contain more epithelial and stromal cells and collagen, which are significant for tumorigenesis as well as tissue-specific progenitor cells that are at risk of transformation to cancer cells.<sup>8,9</sup> Studies have shown that breast density is a strong, modifiable and measurable risk factor for breast cancer.<sup>10–13</sup> Additionally, the masking effect from breast density reduces the performance of screening mammography and limits early detection and treatment of breast cancer.<sup>14</sup> Encouragingly, breast density is reducible, and its reduction has been shown to mitigate breast cancer risk.<sup>13</sup> Therefore, mammographic breast density measurement can be used for breast cancer risk prediction and personalization of breast cancer prevention and control strategies, such as the selection of females who may require breast density reduction interventions. It may also be used for selection of more appropriate imaging pathways for earlier detection of breast cancer.<sup>5,13</sup> Utilization of breast density for these

purposes requires robust and consistent methods for its assessment.

Breast density depicted by the radio-opaque areas on a mammogram can be assessed using qualitative and quantitative (semi-automated and automated) methods.<sup>15–17</sup> Qualitative methods assign breast density grades based on visual assessment of the relative proportions of dense tissue, fat and prominence of ducts and include breast imaging reporting and data system (BI-RADS®), visual analogue scale and Wolfe, Tabar and Boyd assessment methods.<sup>15,18,19</sup> Semi-automated methods use segmentation and thresholding techniques to quantify the percentage of dense tissue on a mammogram and include planimetry and interactive thresholding methods such as Cumulus and Madena.<sup>20,21</sup> Automated methods use mathematical, statistical and physical modelling to calculate breast density; such automated methods include computerized texture-based techniques, calibration approaches and dual X-ray absorptiometry.<sup>22–24</sup> Others are automated thresholding approaches, such as Autodensity and MedDensity,<sup>25,26</sup> and three physical model-based techniques: standard mammographic form (SMF), Volpara and Quantra.<sup>27–29</sup> Irrespective of the method of measurement, breast density has been shown to be a potent risk factor for breast cancer.

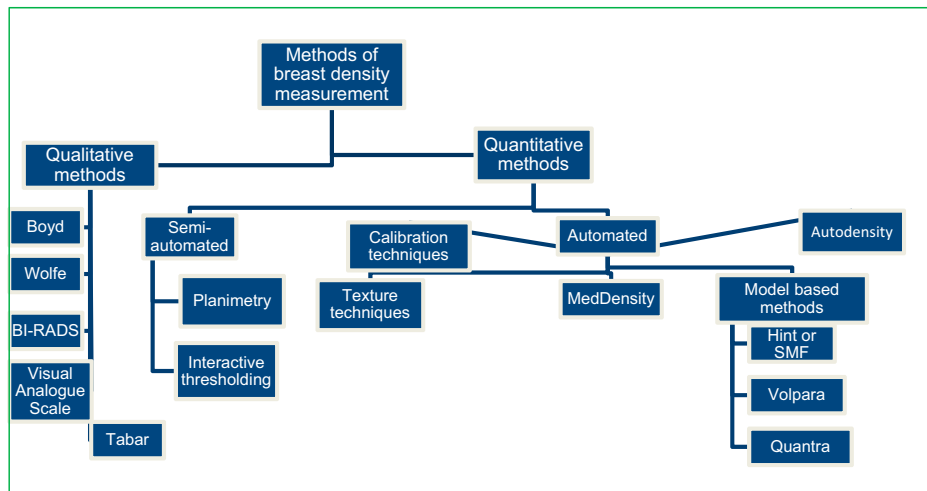
Figure 1. Principles of digital breast tomosynthesis: (a) tube rotations relative to the detector and (b) acquired image slices. Image courtesy of Hologic Inc.; Bedford, MA © 2011. All rights reserved.



Many studies on mammographic breast density measurement are based on film–screen mammography and digital mammography (DM), which produce two-dimensional (2D) images of a three-dimensional (3D) breast. Qualitative methods have been shown to be poorly reproducible with these modalities; they have wide inter-reader agreement with Kappa (*k*) values ranging from 0.37 to 0.91.<sup>26,30</sup> Quantitative methods have better reproducibility with these modalities; however, there are concerns that quantitative area measurement of breast density as percentage mammographic density (PMD) is not representative of the tissue

at risk of breast cancer, and that it is more reasonable to measure the volume of only the fibroglandular tissue, which is more related to the dense tissue at risk instead of PMD.<sup>16,31</sup> Another concern is that volumetric breast density measurement with 2D mammography is limited owing to the absence of depth information in such mammograms;<sup>31</sup> methods estimating mammographic breast density with 2D mammography attempt to take into account variation in breast tissue thickness by modelling; however, with all models, there are assumptions made that may not be necessarily correct for an individual patient.

Figure 2. Methods of breast density measurement. BI-RADS®, breast imaging reporting and data systems; SMF, standard mammographic form.



Digital breast tomosynthesis (DBT) has gained acceptance as a tool for imaging of the symptomatic breast and as an adjunct to DM in screening.<sup>32,33</sup> Breast density assessment with DBT needs some consideration since quantitative methods are modelled for 2D mammography. DBT is a 3D imaging modality utilizing the concept of conventional tomography but a limited angle of tube movement (11–60°) to acquire depth information from the breast (Figure 1a,b).<sup>34</sup> With the removal of anatomical noise (superimposed skin and subcutaneous tissue) in DBT images, quantitatively assessed breast density is expected to be lower than DM. On the other hand, more dense tissue becomes apparent to a subjective reader and qualitatively assessed breast density with DBT is expected to be higher relative to DM. It is therefore important to have a standardized robust, reliable and reproducible assessment method to avoid variability in breast density measurement as this will impact on clinical decision-making for females undergoing breast screening. There are several contending methods (Figure 2), each of which has its own merits; this review briefly examines the links between breast density and breast cancer. It also examines methods that have been used for measurement of mammographic breast density with DBT to ascertain which can be considered the best approach.

## METHODS AND MATERIALS

### Search strategy

The preferred reporting items for systematic reviews and meta analysis (PRISMA) search strategy was employed to search for articles in MEDLINE, EMBASE, CINAHL (EbSCOhost), PubMed, SPIE library, Cochrane library, Web of Science and Scopus databases. In order to access more information, we conducted a Google search, and reference lists of published articles were examined to identify additional articles not identified in the database searches. Searches were conducted using the following terms: breast density assessment with digital breast tomosynthesis, breast density and digital breast tomosynthesis, methods for breast density assessment, and breast density and breast cancer risk. The PICOS system was used to evaluate each article for relevance (Table 1).

### Inclusion criteria

Articles were included if they described breast density and breast density measurement with DBT and were published in English language from January 2000 to March 2014. Articles were qualitatively assessed for study quality and risk of bias to ensure that they fit the inclusion criteria. Articles that did not fulfil these criteria were excluded.

### Data synthesis

Data extraction was performed independently and blindly by two reviewers with differences of opinion resolved by discussion. Where a consensus was not reached, articles were excluded. The selection was strongly influenced by the guidelines for assessing study quality and risk of bias.<sup>35</sup> Each study was scored high or low by each reviewer; this enabled us to appraise the conduct of such research. Only articles rated high were included in this review.

### Results

The search strategy identified 812 articles published from 2000 to March 2014. Out of these, 11 studies fulfilled the inclusion

Table 1. Criteria for determining study eligibility.

Characteristics	Criteria
Study year	Studies published from January 2000 to March 2014
Study type	1. Case-control trials 2. Cohort studies
Population	Females of all ages
Intervention	1. Breast density measurement methods
Comparator	1. Robustness, reliability and reproducibility of density measurement methods
Outcomes	1. Mammographic breast density measurement performance 2. Breast cancer risk prediction capability

Studies were characterized according to the year of publication, study type, population, intervention, comparators and outcomes.

criteria. All studies were rated high and were used to assess the performance of breast density assessment methods with DBT. A total of 842 cases (mammograms) were evaluated by these studies.

### Breast density and breast cancer risk

Studies have shown that breasts with >75% dense tissue have a four- to six-fold higher risk of developing breast cancer relative to those with <10% dense tissue.<sup>4,5,15</sup> The high risk of breast cancer from breast density has been linked to high epithelial and stromal cells and collagen concentrations in the microenvironment of dense breasts<sup>8,36</sup> and increased activity of mitogens and mutagens in dense breast tissue.<sup>37,38</sup> Breast cancers evolve from epithelial cells; stromal cells stimulate epithelial cells' proliferation through insulin-like growth factor 1 (IGF-1) and transforming growth factor beta,<sup>39</sup> and collagen in the breast microenvironment assists in tumour reorganization.<sup>40</sup> Therefore, increased concentration of each of these components increases the risk of carcinogenesis. Similarly, biological interaction among these three components results in stretching and stiffening of each component, initiating processes that lead to cancer.<sup>36,40</sup> Additionally, dense tissue contains high concentrations of mitogens, such as IGF-1 and oestrogen,<sup>41,42</sup> and mutagens such as cytochrome P450 1A2.<sup>37</sup> Therefore, there is increased exposure of proliferating progenitor epithelial and stromal cells in dense breasts to the toxic metabolites of mitogens and mutagens, increasing the probability of their transformation to cancer.<sup>43</sup> Thus, it is clear that measuring breast density can give an indication of breast cancer risk, which may allow for earlier adoption of preventive and control measures.

### Qualitative methods of mammographic breast density assessment

Qualitative methods involve subjective decision-making and grading of mammographic breast density based on visual perception. Such methods include Wolfe, Tabar and Boyd assessment methods, as well as visual analogue scale and BI-RADS. These methods have potential applicability for breast density assessment with DBT; however, only BI-RADS has been used with DBT to date.<sup>26,44</sup>

Table 2. Summaries of studies on breast density measurement with digital breast tomosynthesis (DBT)

Author	Study method	Quality rating of study <sup>35</sup>	Number of participants	Results/comparator	Conclusion/outcomes																					
Kontos et al <sup>50</sup>	Cumulus (cohort study)	High	52 females	<p>Correlation between texture features on DBT vs DM, <math>r = 0.73</math>                      Correlation of %BD vs parenchymal features</p> <table border="1"> <thead> <tr> <th>Features</th> <th>DM</th> <th>DBT</th> </tr> </thead> <tbody> <tr> <td>Skewness</td> <td>-0.18 (0.26)</td> <td>0.18 (0.26)</td> </tr> <tr> <td>Coarseness</td> <td>-0.15 (0.34)</td> <td>0.46 (0.003)</td> </tr> <tr> <td>Contrast</td> <td>-0.25 (0.13)</td> <td>0.31 (0.5)</td> </tr> <tr> <td>Energy</td> <td>-0.29 (0.07)</td> <td>0.36 (0.3)</td> </tr> <tr> <td>Homogeneity</td> <td>0.39 (0.01)</td> <td>0.26 (0.11)</td> </tr> <tr> <td>Dimension</td> <td>0.50 (0.001)</td> <td>0.45 (0.004)</td> </tr> </tbody> </table>	Features	DM	DBT	Skewness	-0.18 (0.26)	0.18 (0.26)	Coarseness	-0.15 (0.34)	0.46 (0.003)	Contrast	-0.25 (0.13)	0.31 (0.5)	Energy	-0.29 (0.07)	0.36 (0.3)	Homogeneity	0.39 (0.01)	0.26 (0.11)	Dimension	0.50 (0.001)	0.45 (0.004)	Cumulus is feasible for breast density measurement in DBT, and parenchymal texture features are more related to breast density on DBT than DM
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Kontos et al <sup>48</sup>	Cumulus (cohort study)	High	71 females	<p>Relationship %BD vs parenchymal features</p> <table border="1"> <thead> <tr> <th>Features</th> <th>DBT</th> <th>DM</th> </tr> </thead> <tbody> <tr> <td>Contrast</td> <td><math>r = 0.48</math></td> <td><math>r = 0.26</math></td> </tr> <tr> <td>Energy</td> <td><math>r = -0.47</math></td> <td><math>r = -0.26</math></td> </tr> <tr> <td>Homogeneity</td> <td><math>r = -0.56</math></td> <td><math>r = -0.33</math></td> </tr> </tbody> </table>	Features	DBT	DM	Contrast	$r = 0.48$	$r = 0.26$	Energy	$r = -0.47$	$r = -0.26$	Homogeneity	$r = -0.56$	$r = -0.33$	Cumulus is reliable for breast density estimation in DBT, and parenchymal texture features are more strongly correlated with percentage mammographic density on DBT than DM									
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Ren et al <sup>51</sup>	Quantra (cohort study)	High	15 projection phantom images	<table border="1"> <thead> <tr> <th>Cases</th> <th>1</th> <th>2</th> <th>3</th> <th>4</th> <th>5</th> <th>6</th> </tr> </thead> <tbody> <tr> <td>DBT</td> <td>13.21</td> <td>19.95</td> <td>17.7</td> <td>17.9</td> <td>21.1</td> <td>17.5</td> </tr> <tr> <td>DM</td> <td>13.4</td> <td>18.3</td> <td>17.4</td> <td>16.7</td> <td>19.2</td> <td>16.1</td> </tr> </tbody> </table>	Cases	1	2	3	4	5	6	DBT	13.21	19.95	17.7	17.9	21.1	17.5	DM	13.4	18.3	17.4	16.7	19.2	16.1	Quantra is a promising tool for density estimation in DBT but needs further algorithm optimization
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Bakic et al <sup>20</sup>	Cumulus (cohort study)	High	40 females	<p>Cumulus intrauser correlation for DBT, <math>r = 0.94</math>                      Breast density correlation for DBT vs DM, <math>r = 0.90</math></p>	Cumulus is very reliable for density estimation with DBT, and breast density of DM is strongly correlated with DBT																					
Bakic et al <sup>52</sup>	Cumulus (cohort study)	High	35 females	<p>Cumulus estimated density                      DBT = <math>28 \pm 19\%</math>; DM = <math>36 \pm 20\%</math>                      Correlation DBT vs DM, <math>r = 0.76</math>                      Agreement DBT vs DM, <math>\kappa = 0.56</math></p>	Cumulus is suitable for measuring breast density with DBT. Breast density is lower in DBT than in DM																					

(Continued)

Table 2. (Continued)

Author	Study method	Quality rating of study <sup>35</sup>	Number of participants	Results/comparator	Conclusion/outcomes
Bakic et al <sup>46</sup>	Cumulus (cohort study)	High	39 females	<i>Cumulus interuser correlation</i> DBT, $\rho = 0.85$ DM, $\rho = 0.75$ <i>Breast density with DBT vs DM</i> $\rho = 0.91$ ; $k = 0.79$	Cumulus is suitable for density estimation on DBT, and percentage density estimated from reconstructed and projection images are similar
Tagliafico et al <sup>26</sup>	MedDensity and BI-RADS® (cohort study)	High	50 females	<i>MedDensity vs BI-RADS correlation</i> Four-grade scale (D1–4), $r = 0.54$ Two-grade scale (D1–2 vs D3–4), $r = 0.78$ BI-RADS inter-reader agreement, $k = 0.80$ Breast density with DBT vs DM BI-RADS intrareader agreement, $k = 0.81$ – $0.86$ BI-RADS inter-reader agreement, $k = 0.91$	Moderate breast density correlation between MedDensity and BI-RADS. Density is underestimated on DBT relative to DM
Tagliafico et al <sup>47</sup>	Cumulus and BI-RADS (cohort study)	High	160 females	Correlation between BI-RADS and Cumulus, $r = 0.94$ <i>Cumulus density correlation between modalities</i> MRI vs DBT, $r = 0.95$ DM vs DBT, $r = 0.97$ BI-RADS agreement for DM and DBT, $k = 0.91$	Cumulus and BI-RADS are reliable methods for estimating breast density from DBT, DM and MRI
Tagliafico et al <sup>44</sup>	MedDensity and BI-RADS (cohort study)	High	160 females	<i>MedDensity vs BI-RADS correlation</i> Four-grade scale (D1–4), $r = 0.48$ Two-grade scale (D1–2 vs D3–4), $r = 0.76$ BI-RADS inter-reader agreement, $k = 0.81$ <i>Breast density with DBT vs DM</i> BI-RADS intrareader agreement, $k = 0.79$ – $0.81$ BI-RADS inter-reader agreement, $k = 0.89$	Moderate density correlation between BI-RADS and MedDensity. Breast density is lower on DBT than on DM
Tromans et al <sup>53</sup>	Volpara (cohort study)	High	20 females	Volumetric breast density correlation for DBT vs 2D mammograms, $r = 0.903$ Agreement between BI-RADS and Volpara, $k = 0.953$	Volpara is robust and a promising substitute for BI-RADS
Regini et al <sup>49</sup>	BI-RADS + Quantra for DM (cohort study)	High	200 mammograms	<i>Breast density with DBT vs DM</i> BI-RADS intrareader agreement, $k = 0.96$	There is no variation in breast density assessed with DBT and DM

BI-RADS, breast imaging reporting and data system; DM, digital mammography;  $k$ , weighted kappa agreement;  $\kappa$ , quadratic kappa agreement;  $r$ , correlation coefficient;  $\rho$ , Spearman's correlation.

BI-RADS grades breast density into four categories: D1 (almost entirely fatty breast); D2 (breast with scattered areas of fibroglandular tissues); D3 (heterogeneously dense breast); D4 (extremely dense breast). It considers D3 and D4 as high-grade densities and D1 and D2 as low-grade densities.<sup>45</sup> BI-RADS has demonstrated strong positive intrareader ( $k = 0.79\text{--}0.86$ ) and inter-reader ( $k = 0.65\text{--}0.91$ ) agreement with DBT;<sup>26,44,46,47</sup> reported inter-reader correlation ( $r$ ) ranged from 0.7 to 0.93, with correlations better for D1 and D4 than for D2 and D3 breast density categories.<sup>26,44</sup> BI-RADS has also shown strong positive intrareader ( $k = 0.79\text{--}0.96$ ) and inter-reader ( $k = 0.79\text{--}0.91$ ) breast density agreement between DBT and DM<sup>44,46–49</sup> (Table 2). The reproducibility of BI-RADS is generally poor owing to reader subjectivity in breast density assessment.<sup>30,45</sup> Poor reproducibility could have different implications for breast cancer risk prediction and choices in screening. To reduce variability and provide an objective measurement of breast density, quantitative approaches were developed for breast density evaluation.

#### Quantitative methods of mammographic breast density assessment

Quantitative methods use mathematical, statistical and physical principles to calculate mammographic breast density and are classified into semi-automated and automated methods.

##### *Semi-automated methods*

Semi-automated methods use thresholding and segmentation techniques to perform area measurement of mammographic breast density as percentages and include planimetry and interactive thresholding methods such as Cumulus and Madena. Of the available semi-automated methods, only Cumulus has as yet been used for breast density estimation with DBT (Table 2).

Cumulus (University of Toronto, Canada) employs computerized thresholding and segmentation techniques to select grey levels for density assessment mainly from the central DBT image slice. Two grey levels are usually selected; the first of these separates the pixels in the image representing the breast from the background and sums these pixels to provide a measure of breast area ( $A_B$ ). The second threshold outlines the fibroglandular (dense) breast tissue and sums the pixels over this area to calculate the area of dense tissue ( $A_D$ ). The software calculates PMD as the percentage of the dense tissue and the total breast area ( $\text{PMD} = A_D/A_B \times 100$ ).<sup>20</sup>

Cumulus has demonstrated strong positive intrauser ( $\rho = 0.88$ ,  $r = 0.89\text{--}0.94$ ) and interuser ( $\rho = 0.85$ ) correlations as well as intrauser ( $k = 0.81$ ) and interuser ( $k = 0.56\text{--}0.79$ ) agreement with DBT. The correlation between Cumulus and BI-RADS ranged from 0.54 to 0.94.<sup>20,50,52</sup> Cumulus has shown a strong positive correlation between parenchymal texture features and breast density with DBT ( $r = 0.73$ );<sup>50</sup> such texture features were shown to be more related to mammographic breast density with DBT than with DM. Cumulus estimated mammographic density values are lower on DBT than on DM;<sup>20,50,52</sup> however, the software has been shown to overestimate breast density by 3% with DBT relative to DM in another study.<sup>20</sup> Reported Cumulus-assessed mammographic breast density correlation between

DBT and DM ranged from  $r = 0.76$  to  $0.97$  and  $\rho = 0.78$  to  $0.91$ .<sup>20,47,52</sup> Cumulus is generally thought to be limited by its binomial categorization of pixel into either 100% dense or 100% fat, and area measurement of breast density as percentages ignoring the 3D characteristics of the breast. Additionally, the dependence of the software on user expertise reduces its reproducibility.<sup>54,55</sup> Removing the human from the evaluation completely would obviously be the solution to the issue of subjective variability; therefore, automated methods of breast density evaluation have been developed.

##### *Automated methods*

Automated methods were developed to allow for objective and consistent breast density assessment of mammograms. Such methods include texture-based techniques, calibration approaches, automated thresholding techniques such as Autodensity and MedDensity and three physics model-based techniques: SMF, Volpara and Quantra. Whilst these methods may have potential for mammographic density estimation with DBT, only three of these have as yet been used with DBT mainly on the central slice and include MedDensity, Volpara and Quantra.<sup>26,51–53</sup> MedDensity performs area measurement of breast density as percentages, whereas Volpara and Quantra measure volumetric breast density and grade such densities into BI-RADS categories. Both Volpara and Quantra are extensions of SMF and employ similar principles for volumetric density estimation, but with some differences such as internal calibration as well as thresholds for classifying density into BI-RADS categories.

MedDensity (developed by Giulio Tagliafico, University of Genova, Italy) is a method based on maximum entropy and uses spatial information for automatic thresholding and segmentation of breast into fatty and dense tissue. The software uses the pixel values of the segmented areas to estimate the area of the dense tissue and total breast area. It calculates PMD as the percentage of the area of the dense tissue and the total breast area. Breast density assessed with DBT using this software is moderately positively correlated with BI-RADS breast density measures; reported correlations ( $r$ ) ranged from 0.48 to 0.78, with correlation better on a two-grade (D1–2 vs D3–4) than a four-grade (D1–4) breast density scale.<sup>26,44,56</sup> MedDensity has shown breast density evaluated with DBT to be lower than that for DM by 11.4%, with the level of breast density underestimation with DBT varying according to BI-RADS categories: 16.0%, 11.0%, 3.5% and 18.1% for BI-RADS 1, 2, 3 and 4, respectively.<sup>44</sup> Although MedDensity is automated, quantum and anatomical noise in the image limits its thresholding capability and reduces the reliability of the software.<sup>46</sup>

Volpara (Volpara Solutions, Matakina Company, Wellington, New Zealand) is based on a relative physics model estimating the volume of fibroglandular tissue relative to the entire breast. It measures breast density by finding a reference point of entirely fat in each image and then estimating X-ray attenuation relative to that point for all other points in the image.<sup>57</sup> The software calculates the volume of dense tissue by integrating the thickness of dense tissue at each pixel level values over the image; it computes the volume of the breast by multiplying the area of the breast by the recorded breast thickness.<sup>58</sup> Volpara calculates

average percentage volumetric breast density as a percentage of the volume of fibroglandular tissue and the total volume of the breast. Volpara generates BI-RADS breast density categories by classifying estimated densities into four volumetric density grades (VDGs): VDG 1 (<4.5%); VDG 2 (4.5–7.5%), VDG 3 (>7.5–15.5%) and VDG 4 (>15.5%); these VDGs correspond to BI-RADS 1–4, respectively.<sup>57,58</sup> Although Volpara is modelled to generate objective BI-RADS scores, it generates relatively lower density values than BI-RADS with DM, but such density values have been shown to be strongly positively correlated with BI-RADS categories.<sup>30,53,58</sup> With DBT, Volpara has shown strong positive agreement ( $k$ ) with BI-RADS (0.953) and a volumetric breast density correlation ( $r$ ) between DBT and 2D mammography of 0.903.<sup>53</sup>

Quantra (Hologic Inc., Bedford, MA) uses the physical modelling of mammographic systems as a basis to calculate volumetric breast density.<sup>51</sup> It estimates the thickness of the fibroglandular breast tissue above each pixel in the image, and sums these pixel values to quantify the total volume of fibroglandular tissue in the breast. It also examines the whole silhouette of the imaged breast to estimate the total volume of the breast. The percentage volumetric breast density is then calculated as a percentage of the estimated fibroglandular tissue volume and the total breast volume.<sup>49,59</sup> The software generates BI-RADS breast density categories by classifying the estimated volumetric breast density into four segments: segment 1 (0–<5.20); segment 2 (5.20–<12.6); segment 3 (12.6–<25.7); and segment 4 (25.7–100.0). These segments correspond to BI-RADS 1–4 density grades, respectively. Quantra has been shown to be an accurate<sup>60</sup> and reproducible tool for quantifying breast density with DM.<sup>61,62</sup> To date, there has been no clinical trial on the feasibility of Quantra for breast density estimation with DBT; it has, however, been assessed on phantoms using the

Quantra results of 2D mammograms as a reference and has shown that breast density based on the central slice is 10% higher in DBT than in DM.<sup>51</sup>

The limitations of this review include that it was restricted to studies published in English. Additionally, DBT is a relatively new technology and few studies have assessed breast density with it. With the acceptance of DBT as an auxiliary for screening, it is increasingly likely that clinical breast density assessment will be performed with it in the near future. This review has presented the evidence related to the performance of breast density assessment methods used with DBT, and the relevance of DBT for volumetric breast density assessment.

## CONCLUSION

DBT images contain depth information useful for volumetric breast density estimation, which is more related to the fibroglandular tissue at risk of breast cancer. With DBT, Cumulus has better reproducibility than BI-RADS but still suffers from subjective variability; MedDensity is limited by image noise, and together with Cumulus do not perform volumetric breast density assessment. Volpara and Quantra calculate volumetric breast density; they are robust, reliable and reproducible and are therefore preferred to other methods. Since BI-RADS is the most common clinical methodology, it may be necessary to calibrate Volpara and Quantra so that their thresholds reflect the BI-RADS categories assigned by expert radiologists. Automation and standardization of breast density measurements across sites may provide clinicians with more reliable breast density information and, therefore, more reliable and consistent selection of choices for breast cancer prevention and control. Without this, variations in breast density measurement will lead to unnecessary differences in clinical decision-making for females undergoing breast screening.

## REFERENCES

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