Granulated Rest Frames Outperform Field of View Restrictors on Visual Search Performance

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Dynamic field of view (FOV) restrictors have been successfully used to reduce visually induced motion sickness (VIMS) during continuous viewpoint motion control (virtual travel) in virtual reality (VR). This benefit, however, comes at the cost of losing peripheral awareness during provocative motion. Likewise, the use of visual references that are stable in relation to the physical environment, called rest frames (RFs), has also been shown to reduce discomfort during virtual travel tasks in VR. We propose a new RF-based design called Granulated Rest Frames (GRFs) with a soft-edged circular cutout in the center that leverages the rest frames’ benefits without completely blocking the user’s peripheral view. The GRF design is application-agnostic and does not rely on context-specific RFs, such as commonly used cockpits. We report on a within-subjects experiment with 20 participants. The results suggest that, by strategically applying GRFs during a visual search session in VR, we can achieve better item searching efficiency as compared to restricted FOV. The effect of GRFs on reducing VIMS remains to be determined by future work.

Keywords: human performance, visual search, rest frames, virtual reality, HCI

1 INTRODUCTION

As part of the immersive experience in virtual reality (VR), navigation in the virtual environment (VE) is an essential action along with selection and manipulation. Several application areas, such as the military (Zyda, 2005; [Dataset] Baumann, J., 1993), medicine (Seymour et al., 2002; Seymour, 2008), athletics (Sorrentino et al., 2005), and manufacturing (Kozak et al., 1993) require long-distance or frequent virtual travel¹. However, virtual travel has the common side effect of visually induced motion sickness (VIMS) (Keshavarz et al., 2014; Jerald, 2015), also known as cybersickness (LaViola, 2000), experienced when visual motion conflicts with motion provided by the vestibular system (Irwin, 1881; Reason, 1970; Reason and Brand, 1975; Johnson, 2005). VIMS generally provokes an unpleasant VR experience and is expressed by headaches, stomach awareness, nausea, and disorientation (Johnson, 2005).

As a way to mitigate VIMS, field of view (FOV) reduction techniques, such as dynamic FOV Restrictors (Fernandes and Feiner, 2016), manipulate the FOV to decrease the peripheral visual flow information presented to users during provocative (travel) motion (Figure 1C). However, FOV...

¹In this paper, we consider virtual travel as continuous motion control of the viewpoint, and distinguish it from other forms of virtual travel, such as teleportation.
Restrictors carry the drawback of potentially leading to lower peripheral awareness, which may demand more effort and time to discern the surrounding scene (Jang et al., 2016). There is an open challenge in identifying a way to maintain the benefits from FOV reduction while maximizing the usage of peripheral FOV.

Rest frames (RFs) have been proposed as an alternative to FOV Restrictors, where they were shown to mitigate VIMS by adding virtual elements that remain stable in relation to the physical environment, such as the use of cockpits in many commercial VR games (e.g., [Dataset] Astrowfis Games Ltd, 2018; CCP Games, 2017). However, RFs are application-dependent and, similar to FOV reduction techniques, usually block large parts of the peripheral view (Cao et al., 2018).

Blocking the user’s peripheral view can have critical influences on performance. Previous studies have shown that the performance in search tasks depends mostly on FOV size, where users tend to adjust their search behavior (such as the visual scanning pattern) based on the size of the FOV (Ragan et al., 2015), and small FOVs cause performance deterioration (Hogervorst et al., 2013). Conversely, the human visual system can tolerate occluding noise to a great extent, even when processing time is very brief (Meng and Potter, 2008). The challenge, then, is to explore designs of RFs to leverage amodal completion—the ability humans have to detect objects as a whole even if they are partially occluded (Michotte et al., 1991)—and VIMS mitigation so that RFs’ adverse effects on the peripheral view are minimized.

In this work, we propose Granulated Rest Frames (GRFs), a new RF-based design that adds random noise-like grains to the periphery of the FOV (Figure 1B), in a similar way that happens with restricted FOV. However, rather than blocking much of the peripheral view as traditional RFs, it provides an optimized way to generate occluding noise-like RFs that allows amodal completion.

Given that RFs reduce VIMS to a similar extent as FOV restrictors, we compared GRFs to restricted FOV on visual search, a common task in many VR applications. We evaluated different designs of GRFs, combining Density and Size to establish how they affect visual search performance. Thus, by comparing visual scanning performance (search time and amount of head rotation) under different conditions, we assessed the hypothesis that GRFs could increase peripheral awareness compared to restricted FOV. Our overarching research questions are, “can users have better visual scanning performance under GRFs, as compared to FOV restrictors?,” and “what is the optimal design of GRFs that can degrade user performance the least?” Specifically, we focus on the effects of FOV restrictors and GRFs on search time and the amount of head rotation during an item searching task in virtual reality.

The results provide a deep understanding of the distinction between GRFs and FOV restrictors on visual scanning task performance. Importantly, our results add to the growing body of literature on the generic generation of RFs (Frey et al., 2007; Lin et al., 2002a; Lin et al., 2004).

2 BACKGROUND

In this section, we review related literature on human tolerance to visual noise, and the impact of FOV restrictors and RFs.

2.1 Visually-Induced Motion Sickness

As early as the 1890s, visually induced disturbances have been reported in some devices like Haunted Swing (Wood, 1895) and eyeglasses with inverting prisms (Stratton, 1897). It results in sensations similar to motion sickness: nausea, dizziness, vertigo, and sweating, among other symptoms. In such situations, those
symptoms caused by visual artifacts are referred to as Visually-induced Motion Sickness (VIMS) (McCauley and Sharkey, 1992; Kennedy et al., 2010). Aside from VIMS, it has also been found that visual stimuli can lead to perceived illusory self-motion (referred to as "vection") (Henn et al., 1974; Dichgans and Brandt, 1978). Later work found that detection can also be triggered by auditory cues (Larsson et al., 2004; Valjamae et al., 2005).

The analysis of the associated occurrence of vection and VIMS provides a way to investigate the mechanism of VIMS. Hettinger et al. (1990) suggested that vection was a necessary precondition for VIMS. Nooij et al. (2017) also discovered that VIMS increases with vection strength. They also found that this relation was detected when pooling correlations across all conditions, but not for all conditions considered individually. In a later study, Nooij et al. (2018) hypothesized that strong vection enhanced the velocity storage, a central integrative network involved in motion sickness. However, some work contradicted the positive relation between vection and VIMS (Webb and Griffin, 2003; Weech and Troje, 2017; Weech et al., 2018; Keshavarz et al., 2019). Kuiper et al. (2019) also argued that vection is not a direct cause of VIMS, but a state that relies on other factors to cause VIMS. One striking part of studying the relation between vection and VIMS is the variability of the results in the literature. For example, Palmisano et al. (2018) failed to find a relation between vection and VIMS in a spontaneous postural activity measurement study. However, they affirmed the contribution of vection to VIMS in other work (Palmisano et al., 2017; Risi and Palmisano, 2019). What the existing research suggests is that the real mechanism of VIMS causation is still not entirely understood.

Despite the divergent conclusions regarding the contributors to VIMS, it is clear that vection relates to the visual stimuli, specifically the optical flow patterns induced by visual stimuli (Telford and Frost, 1993; Palmisano et al., 2000; Bubka et al., 2008; Fuji and Seno, 2020). Seya et al. (2014) described the participants felt stronger vection even if facing much smaller and slower-moving optical flow in the background space than in the foreground space. Other studies also have reported that the more distant stimulus causes vection when visual stimuli differ in depth (Ito and Shibata, 2005; Seno et al., 2009). Although there are studies that disagree with the direct correlation between vection and VIMS, considering the incoherent findings in the literature and the potential positive relationship between vection, VIMS, and optical flow, we hypothesize that certain optical flow types elicit vection and may contribute to VIMS. Our approach is to explore static references at the foreground, while maintaining peripheral awareness through the use of a particular implementation of rest frames, which have been demonstrated to alleviate symptoms of VIMS (Cao et al., 2018).

2.2 Peripheral Vision and Visual Search

Previous work has investigated the link between visual search and peripheral awareness. When performing visual searches, people rely on saccades interleaved with periods of fixation more often than smooth eye movements (Collewijn et al., 1988). The angular speed of eye movement can reach up to 600 deg/s during a saccade (Collewijn et al., 1988). Such high retinal speed requires peripheral vision to select potential targets and guide eyes in search tasks efficiently (Erkelens and Hooge, 1996; Rajashekhar et al., 2002). Peripheral vision operates coarsely on patches containing multiple items rather than accurately on individual items (Rosenholtz et al., 2012). Geringswald and Pollmann (2015) confirmed that peripheral vision loss could prevent integrating local configurations with the global display layout. The integration failure led to reduced spatial configuration learning and impaired contextual cueing in visual search.

2.3 Effects of Visual Noise on Visual Perception

Amodal completion is a remarkable characteristic of the human visual system that helps people tolerate visual noise, reconstruct, and recognize partially occluded objects (Gerbin and Salmaso, 1987; Sekuler and Palmer, 1992). In three-dimensional (3D) scenes—those occurring in VEs—Tse (1999) argued that “mergeable” 3D enclosures are crucial elements in amodal completion. Objects with “mergeable” 3D enclosures are partially occluded can be merged by the human visual system when their surfaces have content-related or similar patterns, even if they are geometrically separated. This concept has been validated by other researchers (Anderson et al., 2002; Nanay, 2010; Soska et al., 2010).

2.4 Effects of Field of View on User Experience in Virtual Environments

Several studies in the literature have pointed out the benefits and detriments to large FOVs in VR. Although large FOVs can increase the sense of presence (Howlett, 1990; Prothero et al., 1995), they have also been demonstrated to worsen VIMS (Icx, 1991; Fernandes and Feiner, 2016). A cue conflict generally explains VIMS: the contradictory information received by visual information and vestibular system triggers the nervous system’s reaction to motion sickness (Reason and Brand, 1975). Hence, the larger the FOV, the more apparent a possible conflict might be, which can result in more sickness. It has been observed that people experience more instability, sickness, and presence with larger FOV (Duh et al., 2001; Lin et al., 2002b).

Studies have also found critical influences of FOV over other aspects of the user experience. Low FOV appears to decrease the user’s sense of presence (Arthur and Brooks, 2000), widen the user’s search paths in a visual search task (Cunningham et al., 1995), increase reaction time in shape identification (Robinett and Holloway, 1992), and significantly impair virtual travel performance (Geruschat et al., 1998; Hassan et al., 2002).

2.5 Effects of Rest Frames on VIMS

Several theories have been proposed to explain VIMS. Among them, the Sensory Conflict Theory and the Postural Instability Theory (Irwin, 1881; Reason, 1970; Reason and Brand, 1975; Riccio and Stoffregen, 1991; Johnson, 2005) focus on the role of vestibular systems or postural balance. The Rest Frames Hypothesis (RFH), on the other hand, emphasizes the role of
spatial-perceptual references on the effects of VIMS (Prothero, 1998; Prothero and Parker, 2003).

According to the Rest Frames Hypothesis (Prothero, 1998; Prothero and Parker, 2003), whose essential concepts were first introduced in the work by Steele (1961), when the brain selects rest frames for the body next-step motion, the preferable option to select rest frames is heavily influenced by what is perceived to be the visual background. The reason is that most visual cues in the environment with coherent motion status are in the visual background.

Studies on projection-based systems, such as CAVEs (Cruz-Neira et al., 1993), show that the seams between screens and elements of the real world visible beyond the screens acting as RFs on the foreground may induce lower VIMS (Cruz-Neira et al., 1993; Lin et al., 2004, Lin et al., 2002a). Lin et al. (2004) employed a Virtual Guiding Avatar to alleviate VIMS, where the avatar, a visual cue, provided a relatively stable cue. The results from a revised simulator sickness questionnaire (SSQ) (Kennedy et al., 1993) indicated that a Virtual Guiding Avatar with rotational cues alone or with translation could reduce VIMS (Lin et al., 2004). Cao et al. (2018) demonstrated that software-based RFs in HMD systems could also effectively achieve similar impacts.

As a whole, prior research points to a benefit in user comfort of restricting the FOV, under the cost of adverse effects to presence and other user experience factors. Rest frames can mitigate these side effects of FOVs while also reducing discomfort. Based on the human ability to amodal completion, we propose to use Granular RFs that look like visual noise. Our contribution takes advantage of the human visual system’s tolerance to noise in the RF design.

3 GRANULATED REST FRAMES

The design of our proposed GRF technique was inspired by human tolerance to visual noise and the ability to mentally recreate the entire object despite parts of it being occluded by other objects (known as amodal completion, Gerbino and Salmaso, 1987; Sekuler and Palmer, 1992). Our solution uses granulation and random distribution of tiny black circles (grains) as RFs.

With GRFs, we combine the benefits of RFs and FOV restrictors into one visualization technique. RFs are always visible to the user, leveraging the benefits of reducing VIMS, as demonstrated by Cao et al. (2018). Furthermore, the central part of the view is always unrestricted, as is the case with FOV restrictors, (Fernandes and Feiner, 2016). Effectively, our GRF design emulates FOV restrictors without completely obstructing the user’s peripheral view, but, instead, offering stable references in that region of the FOV.

Previously, Cao et al., 2018 demonstrated that RFs effectively alleviate VIMS. However, their design used an application-specific prop—a metal net—as the RF, which, to some extent, restricted the FOV due to the metal net blocking parts of the environment. Notably, some parts of the FOV were continually blocked, making it impossible for a person to reconstruct objects beyond the net with amodal completion. Therefore, we proposed an improved design of RFs, which is application-agnostic. The design ensures that each RF grain can be made small enough to allow human amodal completion, and the random distribution minimizes the possibility of patterns that can continuously block the view. Figure 2B shows an example of the GRFs distribution and how it affects the surrounding environment.

Figure 2 summarizes the design of GRFs. Two parameters control their generation: Size and Density. Size represents the amount of FOV (in degrees) that every single grain (black circle) should cover (Figure 2A), which is measured by:

\[
S_z (\text{meter}) = \frac{\text{Size} (\degree)}{2} \times d (\text{meter}) \sim 0.5 \times \frac{\text{Size} \times \pi}{180} \times (\text{radians}) \times d (\text{meter}),
\]

where \(d\) is the distance between the user’s viewpoint and the GRFs, and \(S_z\) is the linear radius of a single grain. For example, \(\text{Size} = 1\) covers 1 degree of the FOV independently of its distance from the user’s head. Smaller \(\text{Size}\) means more RF granulation and, consequently, lower view obstruction.

Density represents the number of GRFs per degree of FOV (Figure 2C). Its value can range from 0, representing no RF coverage (as in Figure 1A), to \(4/(\pi \times S^2)\), where each degree of FOV is totally covered by RFs, acting as a FOV restrictor (Fernandes and Feiner, 2016, as in Figure 1C). Density is calculated by:

\[
\text{Density (GRFs/degree of FOV)} = \frac{N_g \text{ (particles/ degree of FOV)} \times \pi \times (S_z)^2}{d^2} \sim N_g \times \frac{\text{Size}^2}{d^2},
\]

where \(N_g\) represents the number of grains for each degree of FOV, \(d\) is the distance between the user’s viewpoint and the GRFs, and \(\pi \times S^2\) is the size of each GRFs. In other words, Density is the ratio between the area covered by GRFs and the FOV. The level of Density dominates the randomness and sparsity of RFs. High Density makes RFs less random.

In our implementation, the GRF is composed of separated same-size grains (2D circle). All grains are randomly distributed around the user’s head as a sphere with a radius of 1 m. Based on Size, we first calculated the spherical coordinates of all grains with an even distribution. In other words, there is no overlap or gap between any two grains. Then, according to the Density, for each \(1^\times 1^\) area, we randomly choose precalculated coordinates to render the grains. To avoid the computational load for the applications, all rendered grains are combined as one mesh. Additionally, the combined mesh was triggered and rendered by a free plugin from Unity Asset Store—“VR Tunnelling Pro” ([Dataset] SIGTRAP Ltd, 2019).

By controlling the Size and Density parameters, the GRF design can be more user-friendly and application-based. To allow the fovea to be fully available, we left a soft-edged circular cutout at the center of the user’s FOV, as Figure 1A shows. Same as Fernandes and Feiner (2016), the cutout's opacity linearly increases from completely transparent within the 36° inner FOV, to completely opaque beyond the 80° outer FOV.
4 USER STUDY

We conducted a user study to assess different GRF variations on item searching efficiency, specifically, search time and amount of head rotation. Moreover, we compared the effect of GRFs to that of FOV Restrictors (Fernandes and Feiner, 2016). The study was approved by Duke Campus Institutional Review Board.
4.1 Experimental Design

The experiment followed a within-subject design with repeated measures, where Size and Density are the independent variables (IVs) and Time, Accuracy, and Head Rotations are the dependent variables (DVs). A control condition with FOV restrictors recreating the Fernandes and Feiner (2016) technique was added for comparison with our GRF design.

The levels for each IV were:

- Size: 1S(1°), 4S(4°), 7S(7°);
- Density: 25%, 50%, 75%, 100%.

Note that the condition with 100% density represents FOV restrictors. Both of these are considered control conditions in the experiment. Figure 3 shows the FOV coverage of the 9 unique combinations of GRFs. We encoded the combinations as following: size, density. Thus, the code for each conditions was: 1S 25%, 4S 25%, 7S 25%, 1S 50%, 4S 50%, 7S 50%, 1S 75%, 4S 75%, 7S 75%.

The DVs are detailed as follows (all data were sampled at 50 Hz):

- Search Time: Time in seconds to identify a target, calculated by:
  \[
  t = \frac{\text{total time (sec)}}{\text{#identified targets}}
  \]
  \[\text{(3)}\]

- Accuracy: The ratio between the targets counted by the participant and the real number of targets.
- Head rotation: The average amount of head rotations, in degrees, to identify one target

\[
\theta_{\text{head}} = \arccos\left(\frac{P_1 \cdot P_2}{|P_1||P_2|}\right)
\]

where \(P_1\) and \(P_2\) are the unit direction vectors of the user’s face at two different moments. Moreover, \(P_1\) and \(P_2\) are represented as a vector \(\vec{P} = (P_x, P_y, P_z)\). The coordinates were calculated by:

\[
P_x = \sin\psi\cos\theta, P_y = \sin\psi\sin\theta - \cos\psi\sin\phi\cos\theta, P_z = \cos\psi\cos\theta,
\]

where \(\psi, \phi, \theta\) are the angles of head roll, pitch, and yaw.

4.2 Participants

Twenty participants from the same institution participated in the experiment (6 females), with mean ages of 26.75 ± 1.83. They were recruited by posts and emails. All subjects attended the whole session. The subjects read and agreed to an Informed Consent Form before the experiment. Only three of them had little VR experience before the study. Participation was voluntary without compensation.

4.3 Hypotheses

According to the RFH (Prothero, 1998) and its influence on peripheral awareness, we formulated four hypotheses for our study.

- H1: GRFs lead to faster search and lower head rotations as compared to FOV Restrictors.
- H2: Small grains lead to better performance than large grains.
- H3: Low-density GRFs lead to better performance than high-density GRFs.
- H4: Small grains in low-density GRFs are the optimal combination for the lowest search time and head rotation amount.

In order to make our design comparable with dynamic FOV restrictors, we configured the parameters of the circular cutouts with the same inside FOV (IFOV) (36°) and outside FOV (OFOV) (80°) in the center of the FOVs as Fernandes and Feiner (2016). Figure 1A shows an example view of an unrestricted VE; Figure 1B shows a view from a VE with GRFs; Figure 1C shows a VE with the restricted FOV. Red and green circles at the top icon represent IFOV and OFOV, respectively.

4.4 Task and Stimuli

The visual search task is one of the most popular methods to investigate the attention spent in visual object recognition, in which an observer actively scans the visual environment for a particular target among an array of distractors (Treisman and Gelade, 1980; McElree and Carrasco, 1999). The visual search contains feature search and conjunction search. Searching for a target amongst distractors that differ from the target by a simple visual feature, such as color, shape, or orientation, is known as feature search (Treisman, 1998; McElree and Carrasco, 1999). The other one requires a process to distinguish targets from distractors possessing one or more common visual features, which involves bottom-up processes at an early stage to locate analogs to the target fast, and top-down process in later stages to eliminate distractors (McElree and Carrasco, 1999; Shen et al., 2003). It’s believed that visual search in the real world is more similar to conjunction search with less complexity (Alexander and Zelinsky, 2011, 2012; Hout and Goldinger, 2015).

However, either Fernandes and Feiner (2016) or us apply the visual modification to the periphery, which yields more impacts on the bottom-up process of fast locating targets, like feature search, rather than the top-down process that requires one’s previous knowledge (Zhaoping and Frith, 2011; Rosenholtz et al., 2012). Therefore, we designed a classic feature search experiment in VR. The task consisted of identifying and counting targets that were distributed in the distractors. We separated the field into 2 × 2 quadrants for each trial, and each quadrant had a similar number of sparsely and pseudo-randomly distributed targets. The targets were red circles, and the distractors were green triangles with the same size, as seen in Figure 4. We minimized any unintended distractions during the search task by using a blank background.

4.5 Formative Pilot Study

We ran a formative pilot study to determine the stimuli distribution layout around the user. We assessed three layouts: spherical target distribution (360° vertically and horizontally), semi-spherical target distribution (180° vertically and...
The formative study aims to find the layout to yield the largest different search time between the restricted FOV and our design. Results (Table 1) from the formative study show that the conditions that caused a variation in search time were the semi-sphere distribution layout and the complete sphere distribution layout. Thus, we chose the semi-sphere distribution layout since navigation rarely requires the users to look back when they are moving. The final task layout consisted of 8–13 targets (red circles) embedded among 187–192 distractors (half is circles, and the other half is triangles). If the distractor amount is odd, the extra one is a circle.) that were equally distributed along the 3D regions around the user. The target distribution is illustrated in Figure 4.

4.6 Equipment
An Oculus Rift CV1 (about 80° horizontally and about 90° vertically) was used. It’s driven by OculusClient 0.1.0.0 on an Intel(R) Core i7-8700K CPU (3.7 GHz) with 16 GB RAM and an Nvidia GeForce GTX 1080 Ti running Windows 10. 6DOF head tracking allowed the system to render the RF stable relative to the real world even as the user freely moved his head while seated or standing (Figure 4). An XBox One Gamepad wireless controller was used to capture the user’s inputs.

4.7 Procedure
The participants answered a biographical questionnaire and were instructed on the task goals and guided through the XBox controller’s input commands. They practiced in a scene without visual restrictions to get familiar with the task. Then, they experienced 10 conditions in a random sequence. One with fixed FOV restrictor, and 9 conditions with different GRF combinations of Size and Density. Each condition had five trials, followed by a 30-second rest, where the screen of the headset was blacked out, and participants could rest. For each trial, the number of targets randomly ranged from 8 to 13 to avoid a learning effect. When the participant finished the counting, they pressed the Xbox controller right trigger to finish the trial and stop the timer. The targets were deactivated, the time and head rotations data were automatically recorded, and the participant verbally informed the researcher about the number of identified targets and the answer was recorded by the researcher. We instructed the participants to be as accurate as possible, considering time. After the study, the participants were asked their preference on the GRFs and FOV restrictors via the post-experiment questionnaire. The question was “Among all 10 conditions, which one did you feel much easier and more comfortable to identify the target?”

In summary, the experiment consisted of: (3 Size × 3 Density + FOV Restrictors) × 20 participants × 5 Trials = 1,000 unique observations.

5 RESULTS
5.1 Data Analysis
We analyzed whether the Granulated Rest Frames (GRFs) IVs (Size and Density) and their interaction had a significant effect in the Search Time, Accuracy and Amount of Head Rotation with a 2-way repeated measure ANOVA test. We verified if the ANOVA assumption of normality of the residuals was violated with the Shapiro-Wilk test. Then, we used repeated-measures ANOVA to analyze the effect of the nine GRF combinations plus the FOV Restrictors (Fernandes and Feiner, 2016) conditions on Search Time and Amount of Head Rotation, and the non-parametric Friedman test to analyze the effect of the GRF variations plus and the FOV Restrictors condition on Accuracy. The alpha significance level was set to 0.05. We conducted a Post-hoc analysis if a variable was found statistically significant.

5.2 Search Time
We conducted a two-way ANOVA that examined the effect of Size and Density level. There was a statistically significant
interaction between the effect of Size and Density on search time \((F_{4,76} = 2.70, p < 0.037, \eta_p^2 = 0.01)\), while there was no significant difference among the levels Size \((F_{2,38} = 0.24, p = 0.79)\) and Density \((F_{2,38} = 0.10, p = 0.90)\) individually. Simple main effect analysis only showed that Size 1S was significantly faster than Size 4S when Density was 75\% \((p < 0.005)\). Figure 5 shows the interaction between the Size and Density factors.

A single factor analysis was conducted that examined the effect of FOV Restrictors and Rest Frames (9 GRFs variations) conditions. The statistical analysis indicates that there was a statistically significant effect of the conditions on search time \((F_{9,171} = 5.24, p < 0.001, \eta_p^2 = 0.061)\). Post-hoc comparisons using dependent t-tests with Bonferroni correction indicate that the GRFs variations 1S_50\% \((p = 0.029)\), 1S_75\% \((p < 0.001)\), 4S_25\% \((p < 0.001)\), 4S_50\% \((p = 0.013)\), 7S_25\% \((p < 0.013)\) and 7S_50\% \((p = 0.006)\) performed significantly faster than the FOV Restrictors condition. Figure 6 shows the search time results for each condition.

In conclusion, regarding the search time per identified target, the results confirm our first hypothesis that GRFs help users obtain better performance compared to FOV Restrictors. Nevertheless, it only partially demonstrated our second and fourth assumptions that low Size only prevails large Size with large Density.

### 5.3 Accuracy
The Shapiro-Wilk test \((p < 0.05)\), the visual inspection of histograms, normal Q-Q plots, and box plots showed that accuracy was not normally distributed. We then conducted a non-parametric Friedman test to determine the effect of FOV Restrictors and Granulated Rest Frames (9 GRF variations) conditions on target identification accuracy. There was a statistically significant difference in Accuracy depending on Condition, \(\chi^2(9) = 23.45, p = 0.005\). The Wilcoxon post-hoc analysis comparing the levels of FOV Restrictor and GRFs indicates that the GRF 1S_50\% \((M = 0.986 \pm 0.024, p = 0.003)\) and 4S_25\% \((M = 0.984 \pm 0.025, p = 0.021)\) had significantly higher accuracy than the FOV Restrictors \((M = 0.951 \pm 0.076)\).

### 5.4 Amount of Head Rotations
A two-way ANOVA indicates a statistically significant interaction between the effect of Size and Density on search movements \((F_{4,76} = 2.60, p = 0.042, \eta_p^2 = 0.01)\). However, simple main effect analysis did not yield a significant difference among Size and Density combinations.

The single factor analysis examining the effect of FOV Restrictors and Granulated Rest Frames(9 Rest Frames variations) conditions on search movements revealed a statistically significant effect of the conditions on search

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**TABLE 1 | Average search time in the formative study.**

| Layout                        | FOV restrictor | 1S_25% | 1S_75% | 7S_25% | 7S_75% |
|------------------------------|----------------|--------|--------|--------|--------|
| Spherical distribution       | 25.85 s        | 17.61 s| 21.46 s| 19.02 s| 17.06 s|
| Semi-spherical distribution  | 13.28 s        | 10.79 s| 9.06 s | 7.99 s | 8.84 s |
| Quarter spherical distribution| 8.10 s         | 6.98 s | 5.82 s | 6.18 s | 6.20 s |

**FIGURE 5 | Search time for each level of GRF Size and Density. When Density is 75\%, GRFs with Size 1S \((M = 1.236 \pm 0.692)\) perform significantly faster than Size 2S \((M = 1.426 \pm 0.690)\). The circle outlines represent the mean values for each condition.**
movements ($F_{9,171} = 5.76, p < 0.001, \eta^2 = 0.061$). Figure 7 shows the results of search movements. Post-hoc pairwise t-tests with Bonferroni correction (which multiplies p-values by the number of comparisons) comparing the levels of FOV Restrictors and GRFs indicates that the GRF variations 1S_25% ($p = 0.04$), 1S_75% ($p = 0.001$), 4S_25% ($p < 0.001$), 4S_50% ($p = 0.05$), 7S_25% ($p < 0.02$) and 7S_50% ($p = 0.007$) had significantly lower search movements than the FOV Restrictors condition.

As with direct association with search time, the number of head rotations presented highly similar results to search time, where 1S_75%, 4S_25%, 4S_50%, 7S_25% and 7S_50% produced significantly fewer search movements than the FOV Restrictors condition. The only exception is the conditions 1S_25% and 1S_50%, where the former presented fewer head rotations than the reference condition, while under the 1S_50% condition, the search time was faster, even without a difference in head
rotations. Likewise, the results confirmed the first hypothesis, and partially accepted the second and fourth hypotheses, but rejected the third.

5.5 Post-experiment Users Preference
We also collected the participants’ post-experiment preferences on the GRFs and FOV restrictors.

Among all 20 participants, 5 people preferred FOV restrictors, since GRFs distracted them. The other 15 participants preferred GRFs. They reported that it helped them concentrate on the target and use peripheral vision to search for items. “The viewport with GRFs helps in concentration while I can perceive the background as well.” (Subject 1 and 4). Subject 17 mentioned that with the FOV restrictors, it was hard to track the targets due to the limited view. She had to scan the whole graph line by line.

Of the subjects that preferred GRFs, eleven voted against the highest Density. According to their opinions, high Density blocked too much view, and they felt it difficult to see information on the peripheral vision. Size had the same results and responses.

6 DISCUSSION
Based on our understanding of peripheral awareness, visual searching, and previous work (Xiao and Benko, 2016), we hypothesized the granulated rest frames could accelerate people’s visual searching efficiency by contrast to FOV Restrictors (Fernandes and Feiner, 2016) due to it blocks the fewer peripheral view, and our visual systems can reconstruct the environment with amodal completion. As a result of this, we proposed four hypotheses regarding the influence of Size and Density of granulated rest frames over visual search performance.

We reported the results of a user study that compared combinations of size and density and the performance of each GRFs combination with FOV Restrictors. FOV Restrictors are considered the standard and have already been implemented in various applications that require virtual navigation. However, the reduction of FOV has a major limitation in peripheral awareness. As hypothesized, users have significantly better visual perception with GRFs compared to restricted FOV. Both the search time efficiency, accuracy and scene scanning (head rotations) are significantly lower when using the specific settings of size and density. Interestingly, the combinations 1S_75%, 4S_25%, 4S_50%, 7S_25% and 7S_50% achieved better time efficiency and lower head rotations. We analyzed the correlation of time and head rotations using the Pearson test and found a strong effect between the two factors ($r = 0.86, p < 0.0001$), where fewer head rotations led to lower search times. The setting 4S_25% achieves significantly better results for the three DVs tested (Search time, Accuracy and Scene scanning) when compared with FOV Restrictors. The reason for the better performance of GRFs could be due to the improved peripheral awareness and the better distinction between the central vision and peripheral vision. Participants may change the direction they gazed, which can explain the performance improvement. However, we assume the gaze is not a major contributor, due to the participants tending to rotate their head frequently to count the targets during the task. Furthermore, all targets were distributed sparsely, making it difficult to identify the targets just with the gaze.

The significantly faster search time with the combination of Size 1S and Density 75% compared to Size 4S and Density 75%, partially demonstrated our second and fourth hypotheses that only when the Density is very large, the small Size is better than large Size. Along with the fact that no significant main effect but significant interaction effect of Density or Size over the head rotation amount, we can conclude the GRFs’ influence is due to the combination of Density and Size. It might be due to the Density rather than Size causes a more continuous block of FOV. When the Density is low, regardless of the Size, it serves more like a regular FOV containing a more continuously perceived visible peripheral view, which weakens the utilization of amodal completion. That is why only medium Density associated with all Sizes achieved better performance than other conditions.

Therefore, if we want to take advantage of this technique with very high density, it would be best to use small grain sizes. Medium level density can work well with all sizes, but the low-level density performs worse with small grains concerning search time.

One concern we had was the user acceptance of the visual noise added by our technique. The post-experiment questionnaire results suggest that the participants quickly adapted to the GRFs in the peripheral vision, confirming that our implementation of RFs can be successfully tolerated without degrading the experience.

Overall, GRFs helped subjects achieve better peripheral awareness than restricted FOV. On the other hand, performance did not change significantly among conditions with different settings of GRFs. While we can’t make a strong conclusion on the lack of significance, this may suggest that different layouts of GRFs do not influence peripheral awareness. Considering the majority of participants voted against high Density, as well as considering the computational performance, we suggest choosing low-level density combined with medium size to set up the Granulated Rest Frames. Not only because the fundamentals of GRFs are maintaining the advantages of FOV modification, but also using the ability to amodal completion to avoid the loss of peripheral information.

Inspired by the Rest Frames Hypothesis (Prothero, 1998; Prothero and Parker, 2003), previous work applied rest frames in a cockpit design to moderate VIMS (Cao et al., 2018). Similar to Cao et al. (2018), we employed GRFs, a type of foreground visual stimuli that occupies the peripheral FOV, which, according to the Rest Frames Hypothesis, is sufficient to be selected as stationary references, even if there are conflicting cues from the background.

Unlike other types of rest frames, such as a cockpit (Cao et al., 2018) or a virtual nose (Wittinghull et al., 2015), GRFs can be created as visual noise that is generic and content-irrelevant. With Size and Density, the format of GRFs can be more flexible, user-friendly and context-free, which allows this technique to be implemented at the graphics driver-level and seamlessly
applied to various HMDs or applications. However, we still need to validate our proposed GRF design as effective to alleviate VIMS symptoms, which we plan to investigate in future work.

7 CONCLUSION AND FUTURE WORK

We have discussed an improved design of rest frames—Granulated Rest Frames, summarized how we iterated on variations of its design, and performed a within-subject user study exploring seated users’ visual perception performance under granulated rest frames with different parameters to a condition with FOV restrictors. The data from the 20 participants we analyzed indicated that densely granulated rest frames might degrade visual perception efficiency. The comparisons among different conditions suggested users tend to perform more efficiently with GRFs as compared to FOV restrictors.

Spatial recognition ability will be an essential aspect to further exploration, which might be achieved, in part, by employing another counterbalanced study comparing FOV Restrictors and medium-size sparse GRFs. Since we limited our work on the effects of RFs’ peripheral awareness using a static scene, future work should validate the expected benefits of GRFs on VIMS reduction during virtual navigation tasks. Moreover, future improvements of the GRFs technique will require a real-time dynamic modification of the rest frames (size and density), considering the scene features and user locomotion speed and actions. Besides, the search performance highly depends on the specific task and stimulus parameters. For example, when the search task requires top-down processes involving previous knowledge, such as words or elaborate patterns, the performance will likely be different from the task involving bottom-up search processes. Also, the stimulus’s size plays an important role, especially when the GRFs size is similar or larger.

In this study, we investigated the influence of GRFs on bottom-up search processes rather than the top-down processes, which only involves fast locating according to color and shapes. Besides, the GRFs size or density was set at a low level to avoid the impacts. Nevertheless, the potential influence of stimulus size and tasks’ specificity also should be studied in the future.

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

ETHICS STATEMENT

The studies involving human participants were reviewed and approved by Duke Health Institutional Review Board. The patients/participants provided their written informed consent to participate in this study.

AUTHOR CONTRIBUTIONS

ZC performed the study design, examined patients, took samples, and wrote the manuscript draft. JG contributed to the data, crafted the illustrations, corrected the draft of the manuscript. RK contributed to study design, supervised and coordinated the project, and corrected the manuscript. All authors contributed to manuscript revision, read and approved the submitted version.

SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/frvir.2021.604889/full#supplementary-material

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Conflict of Interest: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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