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# BMJ Open Identifying return to work trajectories among employees on sick leave due to mental health problems using latent class transition analysis

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## ABSTRACT

**Objectives** To develop effective return to work (RTW) interventions for employees on sick leave due to mental health problems (MHPs), a better understanding of individual variation in the RTW process is needed. We investigated which RTW trajectories can be identified among employees with MHPs in terms of RTW duration and relapse occurrence during the RTW process. Additionally, we examined how different RTW trajectories can be described in terms of personal and work characteristics.

**Methods** Longitudinal sickness absence registry data were collected retrospectively from the largest Dutch occupational health service. Quantitative RTW information as well as personal and work characteristics were extracted. In total, 9517 employees with a sickness absence due to MHPs were included in the analyses (62 938 data points; RTW durations from 29 to 730 days).

**Results** A latent class transition analysis revealed five distinct RTW trajectories, namely (1) fast RTW with little chance of relapse, (2) slow RTW with little chance of relapse, (3) fast RTW with considerable chance of relapse, (4) slow RTW with considerable chance of relapse and (5) very fast RTW with very small chance of relapse. Differences between employees in the slower and faster trajectories were observed regarding gender, age, type of MHP, organisation sector and organisation size but not regarding part-time work.

**Conclusions** RTW trajectories among employees with MHPs showed large individual variability and differed on personal and work characteristics. Knowledge on different RTW trajectories and their characteristics contributes to the development of personalised RTW treatments, tailored to specific individuals and organisations.

## INTRODUCTION

Mental health problems (MHPs) pose a major challenge for the labour market.<sup>1 2</sup> At any moment, around 20% of the working-age population suffers from MHPs in the average Organisation for Economic Co-operation and Development country.<sup>1-3</sup> MHPs include mental illnesses according to psychiatric classification systems (eg, depression) as well

## Strengths and limitations of this study

- A unique dataset from the largest Dutch occupational health service was used, including sickness absence data from employees with various demographical backgrounds and mental health problems, working for profit and non-profit organisations of various sizes (9517 employees, 62 938 data points).
- The return to work (RTW) process was registered throughout employees' sickness absence period, allowing for a detailed examination of individual variation in RTW trajectories.
- Latent class transition analysis, an innovative and complex data analysis approach, was used to identify distinct trajectories of RTW.
- As our data were originally not gathered for research purposes, information that would be valuable from a research perspective was not always included.

as distress symptoms and burnout.<sup>1</sup> Apart from the individual burden associated with mental ill health, MHPs result in major economic costs for employers and society at large. In fact, MHPs are generally more costly for society than physical diseases such as cancer and diabetes.<sup>4</sup> As opposed to physical diseases, the majority of the societal costs (60%–80%) associated with MHPs are not due to healthcare use but due to increased sickness absence, reduced productivity at work and unemployment.<sup>1</sup> Therefore, it is highly relevant to study return to work (RTW) in this population. The present article focuses on obtaining a better understanding of individual variation in the RTW process, a topic that has received limited attention in earlier research.

In previous research, predictors of work resumption among employees with MHPs as well as interventions that may enhance RTW have been examined. Results of both types of research vary across studies. One systematic



review focusing on depression showed strong evidence for a relation between the duration of a depressive episode and work disability (including longer time to RTW), and moderate evidence for associations between the severity of a depressive episode, comorbidity, previous sickness absence and older age and employee's work disability.<sup>5</sup> Another systematic review focusing on MHPs in general showed strong evidence for a relation between older age and work disability.<sup>6</sup> In both reviews, evidence was limited or inconclusive for many other factors (eg, gender, education, job type, supervisor behaviour).

Regarding interventions, systematic reviews suggest that medication, enhanced primary care and psychotherapeutic interventions aimed at symptom reduction (eg, cognitive behavioural therapy) do not improve RTW among employees on sick leave due to MHPs.<sup>7,8</sup> Psychotherapeutic interventions that include a work-focused component did show encouraging results.<sup>9-11</sup> While work-focused interventions differ per study and generally consist of multiple components,<sup>12,13</sup> successful interventions appear to combine an early, gradual RTW with work-focused cognitive behavioural and/or problem-solving therapy.<sup>12,14</sup> Gradual RTW means that employees resume their work step-by-step in terms of work hours and tasks, until they have fully returned to work. Interestingly, interventions that were effective in terms of RTW did not result in larger reductions in psychological complaints compared with control groups.<sup>9,10</sup> Despite some promising findings, interventions with a work-focused component do not always enhance RTW.<sup>12,14-17</sup> Hence, knowledge to improve RTW among employees with MHPs is still insufficient.

Previous research has mostly focused on RTW as a 'status' isolated from its course (eg, number of days until full RTW). Moreover, previous studies generally treated employees on sick leave due to MHPs as one homogeneous group. Two studies examined individual variation in trajectories of RTW status. These studies showed large heterogeneity in work disability trajectories among employees with MHPs.<sup>18,19</sup> However, little is known about the RTW process and individual variation in this process. While a gradual RTW seems to be a promising component of successful RTW interventions, the implementation of gradual RTW may vary considerably in practice. Employees may return slower or faster and may relapse during their RTW.<sup>12,20,21</sup>

In the present study, we examined which distinct RTW trajectories can be identified among employees on sick leave due to MHPs in terms of RTW duration and relapse occurrence. Additionally, we aimed to provide a description of employees in the different trajectories in terms of personal and work characteristics. A better understanding of different trajectories and their characteristics may allow the development of more effective, personalised RTW interventions. We used a unique dataset from the largest Dutch occupational health service (OHS), including sickness absence data from employees with various demographical backgrounds and MHPs, working

for profit and non-profit organisations of various sizes. In most European countries, sick leave information is deduced from a combination of national registry data on disability benefits and self-report measures. In the Netherlands, the gradual RTW process is well registered by OHS, making it an apt setting for examining RTW trajectories.

## METHODS

### Study context

In the Netherlands, employers have to arrange occupational healthcare for employees within 6 weeks after the start of sick leave. Most employers contract an OHS, an organisation that employs occupational physicians (OPs; qualified medical doctors specialised in occupational health). OPs support employees during their RTW and provide advice to the employer (eg, on necessary work adaptations). The employee and employer make a joint RTW plan within 8 weeks after the start of sick leave, including a potential gradual RTW. For example, an employee may start with 25% of the contract hours and build up this percentage until reaching a full RTW. The RTW process is registered by the OHS. It may be noted that RTW percentage changes reflect the RTW plan and not consultations with the OP. During the first 2 years of sickness absence, employers are required to compensate minimally 70% of an employee's preabsence income. The employer cannot fire the employee during this period (see Dutch Gatekeeper Improvement Act). The National Social Security Institute evaluates the RTW process after 2 years of remaining disability. If the employee is considered incapable of working despite sufficient RTW efforts, the contract with the employer ends and the employee starts receiving unemployment benefits from the government.

### Patient and public involvement

This study builds on a large qualitative study focusing on the perspective of patients and other stakeholders.<sup>22</sup> Furthermore, the study was designed in close collaboration with the OHS. We intend to organise a symposium to share our results with relevant stakeholders.

### Study design

Longitudinal sickness absence data were collected retrospectively from HumanTotalCare, the largest Dutch OHS. Over 80 000 companies nationwide contract this OHS, and their registry has RTW data of 1.5 million employees.

### Sample population

Anonymised sickness absence files of employees who were absent due to an MHP in the year 2014 were extracted from the registry of the OHS (15 580 employees). Employees worked in various sectors for companies of varying sizes. For the classification of MHPs, Dutch OPs use a coding system based on the International Classification of Diseases, version 10. Employees with stress complaints (R45), emotional sleeping disorders (F51.9),

somatoform disorders (F45.0, F45.4, F45.9), adjustment disorders (F43.2, Z73.0), reactions to severe stress (F43.1, F43.9), anxiety disorders (F41.0, F41.1, F40.0, F40.1, F41.9), personality disorders (F60.0, F60.1, F60.2, F60.3, F60.4, F60.6, F60.7, F60.8, F60.9), mood disorders (F30.9, F31.9, F32.9, F34.1, F39), addictions due to psychoactive substances (F10.9, F11.9, F15.9, F19.9), organic psychoses (F09), non-organic psychoses (F20.9, F25.9, F29) and other mental disorders (F48.0, F48.8, F42.9, F44.9, F50.9, F53.9, F63.0, F79, F99) were included. In this coding system, burnout (Z73.0) belongs to the category adjustment disorders.

### Data collection

For each employee on sick leave, the start and end date of the sick leave period, the reporting date and the medical code were indicated in the sickness absence file. Importantly, the RTW percentage at the start of RTW and any changes in the RTW percentage until full RTW (ie, working 100% of contract hours) were included as well. As employers are generally not responsible anymore for an employee's RTW after 2 years of remaining disability, an absence file usually contained RTW information until a maximum of 2 years. Additional data were extracted on personal and work characteristics, including age, gender, contract hours (ie, work hours per week), sector and organisation size.

### Inclusion criteria

Employees with a sickness absence period from 29 to 730 days (ie, 2 years) were selected (13 473 employees). Absences with durations longer than 730 days were deleted as this almost always suggests a registration mistake. For employees who had more than one sickness absence in 2014 (214 employees), the first sickness absence was selected. Furthermore, employees with more than one employer and/or contract (1061 employees) were excluded, as this complicated interpretation. Employees whose contract ended within 7 days after the end of the sickness absence period (1062 employees) were also excluded. For these employees, it was unclear whether they had really returned to work, whether their contract had ended or whether the employer's contract with the OHS had ended. After applying these criteria, 11 350 employees were left in the dataset.

### Data cleaning and preparation

Sickness absence records that contained mistakes were deleted (see online supplementary materials). After data cleaning, the final dataset contained 9517 employees (62 938 data points, ie, RTW percentages, across the sample).

A dataset was created in which the RTW percentage per month was indicated for each employee, until the month of full RTW. If an employee had more than one RTW percentage change during a month, the last RTW percentage was used (this was the RTW percentage that was continued in the next months, until a new RTW percentage change occurred). Furthermore, RTW

percentages were recoded into seven categories: 0%, 1%–19%, 20%–39%, 40%–59%, 60%–79%, 80%–99% and 100%. Next, type of MHP was recoded into four main categories: stress complaints, adjustment disorders, mood disorders and other disorders. Dummy variables were created for the adjustment and mood disorder that occurred most in our dataset, namely burnout and depression. Regarding sector, a distinction was made between profit and non-profit organisations. Dummy variables were created for sectors that were reported more than 500 times in our dataset (industry, wholesale and retail, finance, consultancy, public administration/services, healthcare). Organisation size was recoded into two categories: below 51 employees and above 50 employees.

### Statistical analyses

To identify RTW trajectories, a latent class transition analysis (LCTA) was performed. This analysis allowed the identification of groups of employees showing distinct patterns of RTW (latent classes). We carried out an LCTA as opposed to the more commonly used latent class growth analysis because gradual RTW occurs in stepwise transitions rather than smooth increases. The LCTA consisted of three steps and was carried out in the statistical programme Latent GOLD 5.1 (for more details, see online supplementary materials).<sup>23 24</sup>

In the first step, a latent class transition model was built using the Choice module. RTW percentage category at time  $t$  was used as the 'Choice set' (independent variable), and RTW percentage category at time  $t+1$  as the 'Choice' (dependent variable). Up and down parameters representing the log odds for making an upward or a downward step, respectively, were modelled separately. These parameters indicated how many RTW percentage categories an employee went upward/downward per gradual RTW step. To determine the optimal number of latent classes, we used the Bayesian Information Criterion (BIC). Additionally, a minimum requirement of 5% of the total sample size was used for the smallest latent class.<sup>25</sup> Entropy was reported as a measure of the quality of classification. In the second step, employees were assigned to the different latent classes with a weight that reflected how well each latent class fitted them.<sup>26</sup> In the third step, the latent classes were compared with respect to their composition in terms of personal and work characteristics using a bias-adjusted three-step approach.<sup>27</sup> Wald tests were used to determine whether there were significant differences between the latent classes regarding these characteristics ( $p < 0.05$ ). To avoid discarding potentially interesting relations prematurely in these exploratory analyses, no adjustments for the large sample size or multiple testing were made.

## RESULTS

### Personal and work characteristics

Table 1 shows descriptive statistics for personal and work characteristics. There were somewhat more women than

**Table 1** Descriptive statistics for personal and work characteristics

	M (SD)	Range
Age	41.84 (10.75)	16–66
Contract hours	33.82 (8.09)	1–100
Number of gradual RTW steps	2.35 (1.70)	1–18
	n	%
Gender		
Male	4496	47.3
Female	5019	52.7
Type of MHP (four categories)		
Stress complaints	1530	16.1
Adjustment disorder	5470	57.6
Mood disorder	1157	12.2
Other	1333	14.1
Type of MHP (specific)		
Burnout	889	9.4
Depressive episode	1055	11.1
Sector (two categories)		
Profit	6936	75.9
Non-profit	2197	24.1
Sector (specific)		
Industry	1063	11.6
Wholesale and retail	1861	20.4
Finance	560	6.1
Consultancy	1167	12.8
Public administration/services	518	5.7
Healthcare	991	10.9
Organisation size		
≤50 employees	5022	52.8
>50 employees	4495	47.2

Age: n=9516; hours per week: n=7534; gender: n=9515; type of MHP: n=9490; sector: n=9133. For other characteristics, there are no missing data.

The number of gradual RTW steps was calculated by subtracting one from an employee's total number of RTW percentages. MHP, mental health problem; RTW, return to work.

men in our sample. Furthermore, most employees were diagnosed with an adjustment disorder. The majority of the employees worked in the profit sector, and smaller organisations were somewhat more common than larger organisations. On average, employees resumed their work in two RTW steps.

### Identifying RTW trajectories

LCTAs with one to seven classes were performed (see table 2 for fit statistics). The BIC of the six-class model was slightly lower than the BIC of the five-class model. However, as the sixth class of the six-class model did not meet the minimum requirement of 5% of the total sample

**Table 2** Fit statistics for one to seven class latent transition models (the five-class solution was selected)

Model	LL	BIC	No of parameters	Entropy R <sup>2</sup>	Sample size per class (n)	Sample size per class (proportion of total sample size)
1-class	-77 669.36	155 357.04	2	1.00	9517	1
2-class	-75 409.70	150 865.20	5	0.47	5364/4153	0.56/0.44
3-class	-74 689.30	149 451.89	8	0.44	4247/3388/1882	0.45/0.36/0.20
4-class	-74 258.80	148 618.37	11	0.48	4574/2319/1739/885	0.48/0.24/0.18/0.09
<b>5-class</b>	<b>-74 084.10</b>	<b>148 296.45</b>	<b>14</b>	<b>0.45</b>	<b>4709/1981/1058/908/861</b>	<b>0.49/0.21/0.11/0.10/0.09</b>
6-class	-74 064.55	148 284.84	17	0.45	4635/1999/981/970/859/73	0.49/0.21/0.10/0.10/0.09/0.01
7-class	-74 051.48	148 286.18	20	0.38	3057/2429/1189/974/959/835/74	0.32/0.26/0.12/0.10/0.10/0.09/0.01

Sample size per class: based on most likely class membership.

The selected model is in bold.

BIC, Bayesian Information Criterion; LL, log likelihood.

**Table 3** Up and down parameter estimates per latent class of the chosen five-class model

Parameter	Class 1	Class 2	Class 3	Class 4	Class 5	Wald statistic	P value
	49.5% fast RTW without relapse	20.8% slow RTW without relapse	11.1% fast RTW with relapse	9.5% slow RTW with relapse	9.1% very fast RTW without relapse		
Up	-0.37	-1.01	-0.16	-0.75	7.33	5490.74	<0.001
Down	-7.73	-4.17	-0.41	-0.89	0.91	1045.59	<0.001

Note: Values up parameter coded from 0 to 6. Values down parameter coded from 0 to 5. RTW, return to work.

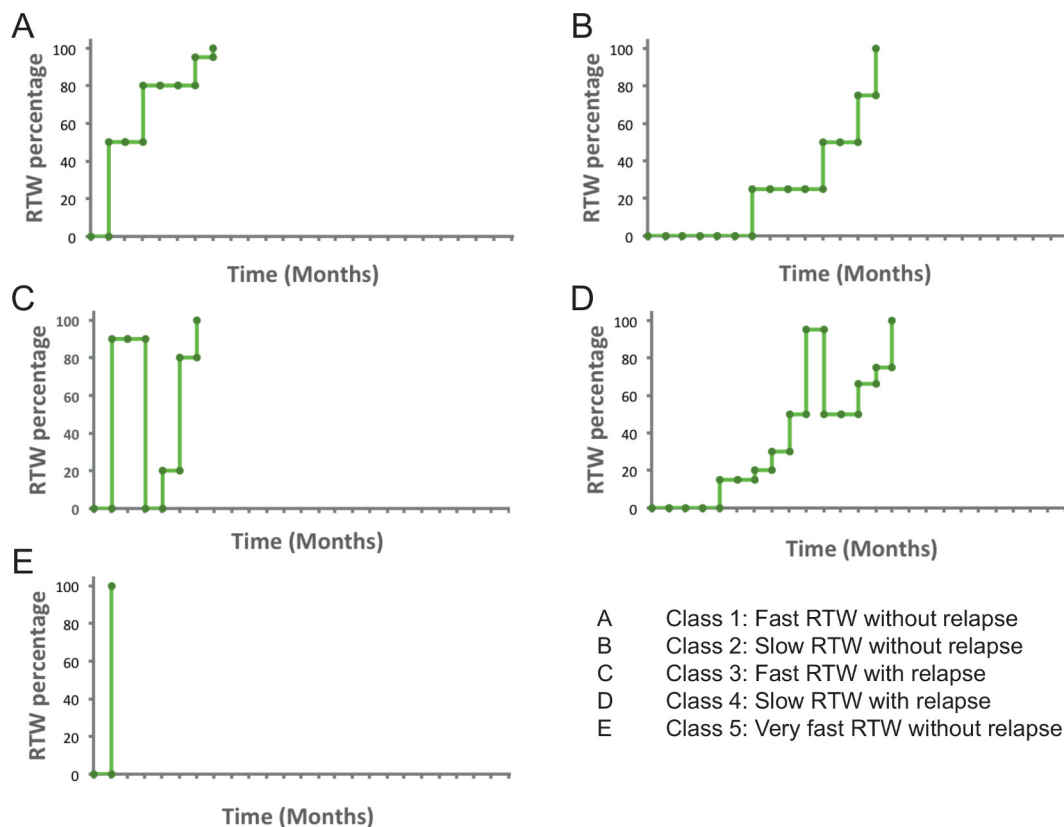
size (this class contained 0.77% of the total sample size), we selected the five-class solution. The six-class model resulted in the same pattern of findings as the five-class model, except that a small group of employees from the fourth class of the five-class model was assigned to a separate, sixth class. This sixth class differed regarding the degree of relapse during the RTW process but not regarding the general pattern of RTW. The five-class model revealed five distinct, meaningful trajectories of RTW.

Table 3 presents up and down parameters of the five-class model. Employees in the first class showed a fast RTW with little chance of relapse during the RTW process (fast RTW without relapse; average of 136 days and 1.96 transitions needed for full RTW). Employees in the second class showed a slow RTW with little chance of relapse (slow RTW without relapse; average of 402 days and 2.47

transitions). In the third class, there was a fast RTW with considerable chance of relapse (fast RTW with relapse; average of 194 days and 3.07 transitions). In the fourth class, there was a slow RTW with considerable chance of relapse (slow RTW with relapse; average of 419 days and 3.54 transitions). Employees in the fifth class returned to work very fast and had a very small chance of relapse (very fast RTW without relapse; average of 49 days and 1.00 transitions). Examples of typical trajectories for each latent class are shown in figure 1 (for transition odds per latent class, see online supplementary materials).

### Characteristics of RTW trajectories

We found significant differences between the five trajectories on age, gender, type of MHP, sector and organisation size (all  $p \leq 0.03$ ). No significant differences were revealed for contract hours (see table 4 for means/



**Figure 1** One example of a typical RTW trajectory per latent class of the five-class model (x-axis shows 24 months). RTW, return to work.

**Table 4** Differences between the five identified RTW trajectories on personal and work characteristics

	Class 1 Fast RTW without relapse		Class 2 Slow RTW without relapse		Class 3 Fast RTW with relapse		Class 4 Slow RTW with relapse		Class 5 Very fast RTW without relapse		Posthoc test (p<0.05)		
	Mean (model based)	Proportion (model based)	Mean (model based)	Proportion (model based)	Mean (model based)	Proportion (model based)	Mean (model based)	Proportion (model based)	Mean (model based)	Proportion (model based)		Wald statistic	P value
Age	41.86	0.48	43.37	0.45	39.75	0.48	42.02	0.40	40.65	0.54	44.62	<0.001	1<2; 1>3, 5 2>3, 5
Hours per week	33.82	0.18	33.83	0.05	34.81	0.19	32.61	0.08	33.75	0.37	8.53	0.07	3<4
Gender (male)	0.48	0.48	0.45	0.45	0.48	0.48	0.40	0.40	0.54	0.54	22.74	<0.001	1>4; 1<5 2<5
Type of MHP (four categories)											397.10	<0.001	4<5
Stress complaints	0.18	0.18	0.05	0.05	0.19	0.19	0.08	0.08	0.37	0.37	250.64	<0.001	1>2, 4; 1<5 2<3, 5
Adjustment disorder	0.62	0.62	0.53	0.53	0.59	0.59	0.54	0.54	0.48	0.48	42.22	<0.001	3>4; 3<5 4<5
Mood disorder	0.08	0.08	0.23	0.23	0.08	0.08	0.21	0.21	0.06	0.06	169.85	<0.001	1>2, 4 2>3, 5
Other	0.12	0.12	0.19	0.19	0.14	0.14	0.17	0.17	0.09	0.09	40.56	<0.001	3<4 4>5
Type of MHP (specific)											1131.05	<0.001	1<2, 4; 1>5 2>3; 2<5 4>5
Burnout	0.08	0.08	0.17	0.17	0.06	0.06	0.12	0.12	0.01	0.01	81.89	<0.001	1<2, 4; 1>5 2>3, 5 3<4; 3>5

Continued

Table 4 Continued

	Proportion (model based)	Proportion (model based)	Proportion (model based)	Proportion (model based)	Proportion (model based)	Wald statistic	P value	Posthoc test (p<0.05)
Depression	0.07	0.21	0.07	0.19	0.06	156.98	<0.001	4>5 1<2, 4 2>3, 5 3<4
Sector (profit)	0.77	0.72	0.78	0.68	0.82	34.51	<0.001	4>5 1>2, 4; 1<5 2<3, 5 3>4
Sector (specific)						108.08	0.001	4<5
Industry	0.13	0.09	0.12	0.09	0.12	10.65	0.03	1>2
Wholesale and retail	0.21	0.18	0.19	0.21	0.22	1.45	0.83	
Finance	0.06	0.06	0.07	0.07	0.05	1.83	0.77	
Consultancy	0.12	0.15	0.15	0.09	0.12	5.56	0.23	
Public administration/ services	0.06	0.08	0.05	0.04	0.01	17.66	0.001	1>5
Healthcare	0.10	0.12	0.10	0.15	0.09	10.60	0.03	1<4 2>5 3>5
Organisation size (≤50 employees)	0.49	0.57	0.53	0.57	0.59	29.02	<0.001	1<2, 4, 5 4>5

Age: n=9516; hours per week: n=7534; gender: n=9515; type of MHP: n=9490; sector: n=9133. For other characteristics, there are no missing data. MHP, mental health problem; RTW, return to work.



proportions per class and significance tests). The most notable differences are summarised below.

### Age

Employees in the slow trajectories (especially class 2) were on average older compared with employees in the fast trajectories (class 1, 3, 5).

### Gender

The proportion of male employees was higher in the fastest trajectory (class 5) compared with one of the fast (class 1) and both of the slow trajectories (class 2, 4).

### Type of MHP

The fast trajectories (class 1, 3, 5) were characterised by more stress complaints compared with the slow trajectories (class 2, 4). Furthermore, the fast trajectories (particularly class 1) contained more employees with adjustment disorders than both the slow trajectories (class 2, 4) and the fastest trajectory (class 5). A different pattern was found when examining burnout specifically. The slow trajectories (class 2, 4) were characterised by more employees with burnout than the fast trajectories (class 1, 3, 5). Finally, employees suffering from mood disorders were more often in the slow trajectories (class 2, 4) than in the fast trajectories (class 1, 3, 5). The same pattern was found for depressive episode.

### Sector

The fast trajectories (class 1, 3, 5) contained more employees working in the profit sector compared with the slow trajectories (class 2, 4). When examining sectors that were reported more than 500 times, differences between the trajectories were found for industry, public administration/services and healthcare, but not for wholesale and retail, finance and consultancy. The proportion of employees from industry was higher in the fast trajectory without relapse (class 1) compared with the slow trajectory without relapse (class 2). Furthermore, the proportion of employees working in public administration/services was lower in the fastest trajectory (class 5) compared with most other trajectories (class 1, 2, 3). The proportion of employees from the healthcare sector was higher in the slow trajectory with relapse (class 4) compared with the fast trajectories without relapse (class 1, 5).

### Organisation size

The proportion of employees who worked in small organisations was lower in the fast RTW without relapse trajectory (class 1) compared with the slow trajectories (class 2, 4) and the fastest trajectory (class 5).

## DISCUSSION

### Summary

We identified five distinct RTW trajectories, namely (1) fast RTW with little chance of relapse, (2) slow RTW with little chance of relapse, (3) fast RTW with considerable chance of relapse, (4) slow RTW with considerable chance

of relapse and (5) very fast RTW with very small chance of relapse. Stress complaints and adjustment disorders were more prevalent in the faster trajectories, while depression and burnout specifically were more prevalent in the slower trajectories. Furthermore, older employees, women and non-profit sector employees showed longer trajectories. Interestingly, part-time employees did not resume work faster than full-time employees. Individual variation in the RTW process among employees with MHPs has received limited attention in earlier research.

### Individual variation in RTW

The identified trajectories varied on RTW duration and relapse occurrence. Around 60% of employees fully resumed work within approximately 6 months (class 1, 3), around 30% of employees within approximately 14 months (class 2, 4) and around 10% of employees within 1/2 months (class 5). About 20% of employees relapsed during the RTW process (class 3, 4). This finding is in line with earlier studies reporting that relapse is common after a partial RTW.<sup>12 20 21</sup> In the short term, faster trajectories without relapse may be most beneficial, especially from an employer perspective. However, slower trajectories with an early onset but a slow RTW process in terms of gradual RTW percentage increases may be more beneficial for certain employees in the long term (eg, employees with more severe MHPs or disadvantageous circumstances at home or at work).

The observed differences between employees in the slower and faster trajectories on personal and work characteristics provide some insight regarding which employees may resume work slowly and what different employees may need in terms of RTW support. Significant statistical differences between trajectories were found for most characteristics. Because of the large sample size, we had a high-powered study, allowing us to detect both smaller and larger differences between trajectories. Especially differences regarding type of MHP seemed clinically relevant. The proportion of employees with stress complaints was higher in the faster trajectories, while the proportion of employees with a mood disorder was higher in the slower trajectories. Previous research also showed some evidence for a relation between depression and longer work disability duration.<sup>6</sup> Furthermore, the proportion of employees with adjustment disorders was higher in the faster trajectories, while the proportion of employees with burnout was higher in the slower trajectories. Work stress models suggest that (chronic) stress may eventually result in more severe MHPs such as burnout and depression.<sup>28</sup> Early intervention for employees with stress complaints and adaptation disorders may prevent more severe complaints and prolonged RTW trajectories.

Regarding other characteristics, employees in the slower trajectories were on average older. This finding is consistent with previous research.<sup>5 6</sup> Furthermore, male employees and profit sector employees were more often in faster trajectories. The differences between men and women may partly be explained by job differences. For

instance, more women work in the healthcare sector, where realising suitable work adaptations may be relatively challenging. Earlier reviews showed limited/inconclusive evidence for relations between gender or sector and work disability.<sup>5,6</sup> Interestingly, part-time employment was not associated with faster work resumption than full-time employment. An explanation may be that people generally work part time for a reason (eg, child care). It may be noted that age, gender and type of MHP are likely to be related. Hence, differences between RTW trajectories in terms of MHPs may partly be explained by differences in age and/or gender.

Interestingly, no differences on personal and work characteristics were observed when comparing trajectories with and without relapse. The finding that these trajectories did not differ on demographical factors, type of MHP or work context characteristics suggests that differences between trajectories with and without relapse may be explained by psychological and human work-related factors (eg, supervisor behaviour, sense of control, general working conditions).<sup>29-32</sup> Importantly, this would imply that relapse depends on factors that can be influenced to a certain extent.

### Strengths and limitations of this study

We used a unique sickness absence dataset from the largest Dutch OHS. Hence, a major strength was the large sample, including employees with various demographical backgrounds, MHPs and work environments. Gradual RTW percentages were registered throughout employees' sickness absence period, allowing for a detailed examination of individual variation in the RTW process. Latent class transition analysis, an innovative and complex data analysis approach, was used to identify distinct trajectories of RTW. Latent class transition analysis is more suitable for data with stepwise changes than the more common latent class growth analysis.

While our data were a rich source of information on the RTW process, these data were not gathered for research purposes. Therefore, information that would be valuable from a research perspective was not always included. For instance, it is unknown whether employees were on sick leave before, there was no information on comorbidity, relevant psychological variables (eg, self-efficacy)<sup>33</sup> were not measured, and it is unknown whether employees participated in interventions during their RTW. Another limitation of our study was that employees with temporary contracts had to be excluded if their contract ended during the sickness absence. Furthermore, OPs generally reported the broader category of a MHP (eg, adjustment disorder) but not always a specific diagnosis (eg, burnout). Finally, it is possible that data deviated from reality at times (eg, an employee's RTW percentage may not have been adapted immediately after a partial RTW) and that information on work characteristics was not always up to date.

### Future research and conclusion

Future research may investigate whether OPs recognise the different RTW trajectories and how knowledge

about different trajectories affects treatment decisions (eg, using focus groups). Since no differences between trajectories with and without relapse were found on the characteristics included in this study, future studies may examine whether trajectories differ on psychological and human work-related characteristics such as RTW self-efficacy,<sup>33</sup> experienced autonomy over the RTW process,<sup>29</sup> frequency and timing of communication between different stakeholders<sup>30,31</sup> and psychosocial work environment.<sup>32</sup> To study this, it is highly relevant for OHS to start recording these factors systematically in collaboration with researchers. Future research may also relate different trajectories to sustainable work resumption, productivity and well-being. This may reveal whether certain trajectories are disadvantageous for some employees in the long term.

In conclusion, using data from the largest Dutch OHS, we found significant heterogeneity in RTW trajectories among employees with MHPs. Early support for employees with stress complaints and adaptation disorders may prevent more severe complaints and prolonged trajectories. The finding that trajectories with and without relapse did not differ on demographical factors, type of MHP and work context characteristics suggests that differences between employees in these trajectories may be explained by (potentially modifiable) psychological and human work-related factors. Knowledge on different RTW trajectories and their characteristics contributes to the development of personalised RTW treatments, tailored to specific individuals and organisations.

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