

Prescribing cascades in community-dwelling adults: A systematic review

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Abstract

The misattribution of an adverse drug reaction (ADR) as a symptom or illness can lead to the prescribing of additional medication, referred to as a prescribing cascade. The aim of this systematic review is to identify published prescribing cascades in community-dwelling adults. A systematic review was reported in line with the PRISMA guidelines and pre-registered with PROSPERO. Electronic databases (Medline [Ovid], EMBASE, PsycINFO, CINAHL, Cochrane Library) and grey literature sources were searched. Inclusion criteria: community-dwelling adults; risk-prescription medication; outcomes-initiation of new medicine to "treat" or reduce ADR risk; study type-cohort, cross-sectional, case-control, and case-series studies. Title/abstract screening, full-text screening, data extraction, and methodological quality assessment were conducted independently in duplicate. A narrative synthesis was conducted. A total of 101 studies (reported in 103 publications) were included. Study sample sizes ranged from 126 to 11 593 989 participants and 15 studies examined older adults specifically (≥ 60 years). Seventy-eight of 101 studies reported a potential prescribing cascade including calcium channel blockers to loop diuretic ($n = 5$), amiodarone to levothyroxine ($n = 5$), inhaled corticosteroid to topical antifungal ($n = 4$), antipsychotic to anti-Parkinson drug ($n = 4$), and acetylcholinesterase inhibitor to urinary incontinence drugs ($n = 4$). Identified prescribing cascades occurred within three months to one year following initial medication. Methodological quality varied across included studies. Prescribing cascades occur for a broad range of medications. ADRs should be included in the differential diagnosis for patients presenting with new symptoms, particularly older adults and those who started a new medication in the preceding 12 months.

KEY WORDS

appropriate prescribing, community-dwelling adults, prescribing cascades, systematic review

Abbreviations: ADR, adverse drug reaction; ATC, Anatomical Therapeutic Classification; CCB, calcium channel blocker; ED, Emergency Departments; PRISMA, Preferred Reporting Items for Systematic Reviews and Meta-analysis; TRIP, Turning Research Into Practice.

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1 | BACKGROUND

A prescribing cascade occurs when a medication is used to treat or prevent the adverse effects of another medication.¹⁻³ An unintentional prescribing cascade occurs when the adverse drug reaction (ADR) is misinterpreted as a new medical condition, leading to the prescription of new medication to treat the emerging symptoms.⁴ For example, calcium channel blocker (CCB) induced lower extremity oedema may be misinterpreted as a sign of congestive heart failure and result in the inappropriate prescribing of a loop diuretic to alleviate the oedema instead of simply switching the CCB to an alternative class antihypertensive agent.⁵⁻⁷ Intentional prescribing cascades occur when the ADR is recognised and a subsequent medication is prescribed to combat this ADR either via treatment of the ADR or prevention of it in the first instance.⁴ Prescribing cascades can be further characterised as either appropriate (potential benefits>risks), or inappropriate (risks>potential benefits).⁴ Furthermore, this characterisation of appropriateness is a dynamic entity; an appropriate prescribing cascade can become inappropriate over time, particularly should the clinical circumstances of the patient change.⁴

It is not clear what drives prescribing cascades. Older adults may be more vulnerable due to the nonspecific nature of ADR symptoms in older adults, e.g. falls, fatigue or constipation, all of which have multiple potential causes.⁸ Multimorbidity, which is more common in older adults, may also make the identification of new onset ADRs more challenging.^{9,10} However, the failure to correctly identify an ADR and the resultant prescribing cascade compounds the risk for medication-related harm.

To date prescribing cascades have remained under-researched. A previous scoping review identified only 10 original investigations and seven case reports that examined prescribing cascades.¹¹ In order to optimise prescribing, it is vital that clinically relevant prescribing cascades that commonly occur in practice are identified. The objective of this systematic review was to identify published prescribing cascades in community-dwelling adults.

2 | MATERIALS AND METHODS

2.1 | Search protocol

The study protocol was previously published¹² and pre-registered with PROSPERO [CRD42021243163].¹³ This study was reported according to the Preferred Reporting Items for Systematic Reviews and Meta-analysis (PRISMA) guidelines.^{14,15} (eTable 1 and eTable 2 in Appendix S1).

2.2 | Search strategy

Searches were conducted in the following databases: Medline (Ovid), EMBASE, PsycInfo, CINAHL and the Cochrane Library. Searches

were initially conducted from inception to March 2021 and updated in February 2022. The search strategy (eBox 1 in Appendix S1) was developed in consultation with an experienced librarian. No restrictions were placed on language or publication year. Grey literature database searches were conducted in MedNar, Dart Europe, Open Grey, and the Turning Research Into Practice (TRIP) databases using keyword searches. Forwards and backwards citation searching of articles selected for full text review was also conducted. Retrieved results were exported to EndNote X9 prior to screening and study selection using Covidence® systematic review management system. Following duplicates removal, titles and abstracts were independently screened by two reviewers (AD and EW, OC or FS) according to inclusion criteria. Disagreements were managed by consensus. Additional information was sought from study authors where necessary.

Studies were included if they met the following criteria:

1. Population: community-dwelling adults (≥ 18 years).
2. Risk: prescription of medication that had the potential to cause an ADR that resulted in the prescription of further medication.
3. Outcome: prescribing cascade defined as the initiation of a new medication to 'treat' an ADR (unintentional cascade) or to reduce the risk of an ADR (intentional cascade).
4. Study type: prospective or retrospective cohort, cross-sectional, case-crossover, case-control or case-series studies.
5. Setting: primary care and community settings, including ambulatory care.

2.2.1 | Exclusion criteria

The following studies were excluded;

1. Population of interest < 18 years;
2. Studies conducted solely in nursing homes, residential care, inpatient settings or Emergency Departments (ED);
3. Case reports

2.3 | Data extraction and quality assessment

Data extraction was conducted by two independent reviewers (AD and EW, OC or FS) using a standardised Microsoft Excel proforma. (see eBox 2, Appendix S1). The methodological quality of included publications was independently performed in duplicate (AD and EW, OC or FS) using the appropriate JBI- Critical Appraisal checklist (eBox 3, Appendix S1). Data synthesis was conducted using a narrative synthesis. Alluvial plots of drug pair combinations were created, using R-Studio 2021.09.2 statistical software using the ggalluvial package, to identify the drug-pair combinations examined and to summarise the overall quantitative association reported.

3 | RESULTS

3.1 | Study identification

The study identification flow diagram is presented in [Figure 1](#). A total of 103 publications relating to 101 studies met the inclusion criteria. Three publications included data from the same study relating to updated data collection time periods (2000–2006; 2000–2010; and 2000–2012).^{16–18} Thus, only the final study publication,¹⁸ which contained the entire data collection period, was included in the narrative synthesis.

3.2 | Study population demographics

Seventy-nine studies presented study participants demographics, of which 15 specifically examined older adults (≥ 60 years), with different age-related thresholds (e.g. ≥ 60 ; ≥ 65 ; ≥ 66 years) used across studies.^{5,19–32} Thirteen studies reported analyses stratified by age.^{7,33–44} Total study sample sizes ranged from 126⁴⁵ to 11 593 989⁴⁶ participants. (See eTable 3, Appendix S1).

3.3 | Methodological approach to analysis

Most studies ($n = 88$) were retrospective cohort studies,^{5,7,18,21,23–27,29–34,36–44,47–113} three of which incorporated a case-control study within the study design^{49,85,110} and one that conducted a preliminary cross-sectional study.¹¹² Five were case-control studies,^{19,20} five cross-sectional studies,^{6,46,114–116} and three

case-crossover studies.^{45,117,118} All studies used routine data (health insurance claims, prescription dispensing, clinical databases, national health surveys and pharmacovigilance data). In total, 83 studies examined dispensed prescriptions whereas 18 studies examined prescribed medications (see eTable 3, Appendix S1).

Of 101 studies, 62 used prescription sequence symmetry analysis (PSSA) to determine the ratio of participants who initiated two medications in both possible sequences (i.e. Drug A → Drug B vs. Drug B → Drug A), with the majority ($n = 52$) adjusting for prescribing trends.

Several studies reported stratified results by dosage,^{5,7,28,29,39} concomitant medication use or polypharmacy,^{7,40,44,85,104} duration,^{32,94} comorbidity,^{36,38,40,44} race³⁴ and nursing home residence.²⁶ For other studies, analyses were adjusted by age,^{20,22,24,30,52,71,89} sex,^{20,22,24,30,52,71,73,82} race,^{22,24} dose,^{52,71} nursing home residence,²² concomitant medication or polypharmacy,^{22,52,71} comorbidity,²⁴ with some studies conducting adjusted analyses but not reporting the independent association of these covariates.^{23,27,31,88,117,118}

Length of follow up ranged from one month^{55,91,107,118} to seven years,¹¹³ with the majority over one year ($n = 33$ studies).

3.4 | Initial medication(s) prescribed to patient

A broad range of medication types were examined as potentially precipitating a prescribing cascade (see [Table 1](#) and column 1, [Figure 2A](#)). Ninety-four studies were hypothesis-driven or examined a predefined list of medications ([Table 2](#) and [Figure 2A](#)). Seven studies conducting exploratory analyses to identify new signals of potential prescribing cascades are not represented in [Fi](#)

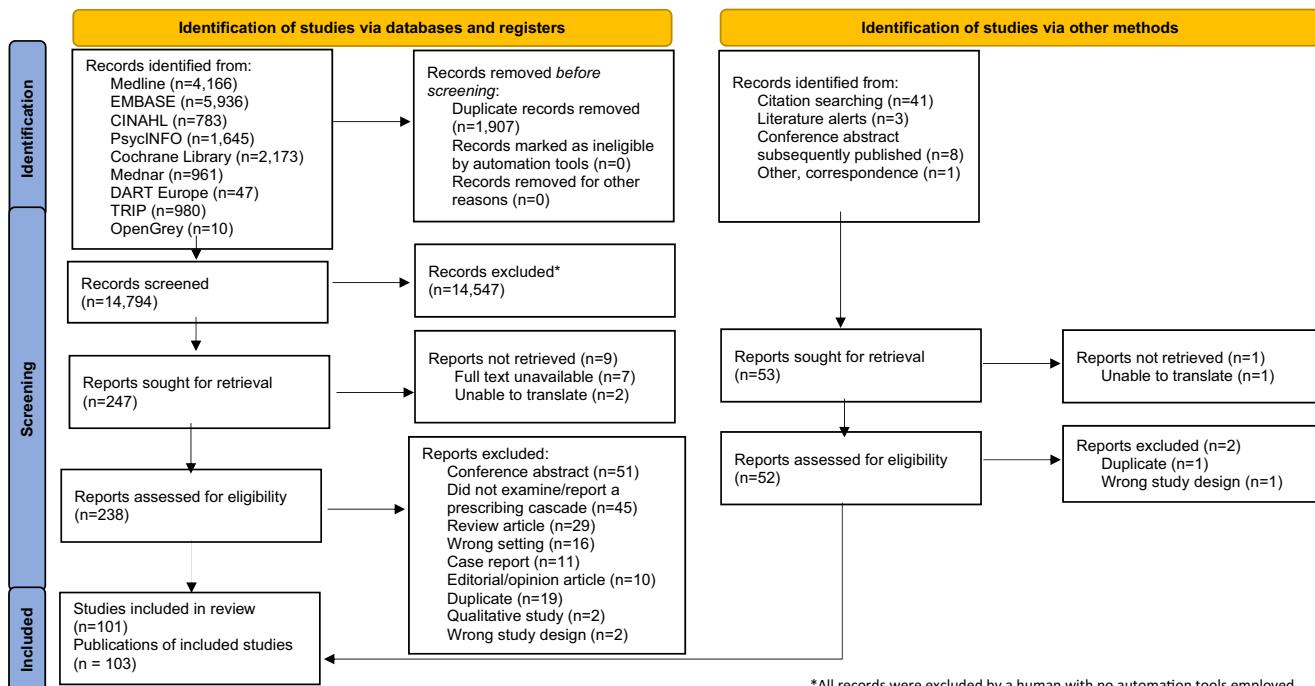


FIGURE 1 PRISMA flow diagram of included studies.

TABLE 1 Primary results of included studies by ATC pharmacological classification ($n = 101$)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Alimentary tract and metabolism				
Adimadhyam (2019) ⁴⁷	Sodium/Glucose cotransporter-2 inhibitors (SGLT2-I)	Genital mycotic infections	Antifungal	PSSA: SGLT2-I → Antifungal ± 365 days aSR 1.24 (95%CI 1.20–1.28)
Avorn (1995) ²⁰	Metoclopramide	Extrapyramidal symptoms (EPS)	Anti-Parkinson drug (APD)	Metoclopramide → APD (<90 days) aOR 3.04 (95%CI 2.22–4.17)
Gadzhanova (2017) ⁸⁸	SGLT2-I Dipeptidyl peptidase 4 inhibitor (DPP4-I)	Urinary or genital infections	Trimethoprim Nitrofurantoin Norfloxacin	Risk of UTI (<6 months) SGLT2-I users (3.6%) compared to DPP4-I users (4.9%), aHR 0.90 (95%CI 0.66–1.24)
Janetzki (2021) ⁹	PPI	Development or exacerbation of chronic obstructive pulmonary disease (COPD)	Long-acting muscarinic antagonist (LAMA) or long-acting beta-2 agonist (LABA) listed for the treatment of COPD	Risk of genital infections (<6 months) SGLT2-I users (2.9%) compared with DPP4-I users (0.9%), aHR 3.50 (95%CI 1.95–5.89)
Lund (2021) ¹¹¹	SGLT2-I Glucagon-like peptide-1 receptor agonists (GLP1-RA)	Gout	Any uric acid lowering therapy, colchicine or first hospital diagnosis of gout (composite)	PSSA: PPI → LAMA/LABA ± 1 year Omeprazole: aSR = 1.29 (95%CI 1.22–1.36) Esomeprazole: aSR = 1.25 (95%CI 1.22–1.29) Rabeprazole: aSR = 1.15 (95%CI 1.08–1.21) Pantoprazole: aSR = 1.08 (95%CI 1.05–1.12) Lansoprazole: aSR = 1.08 (95%CI 0.96–1.22) Risk of gout <3 years: intention to treat analysis HR: 0.58 (0.44 to 0.75) [GLP1-RA as referent] Risk of gout <3 years: per-protocol analysis HR: 0.48 (0.33 to 0.70) [GLP1-RA as referent]
Park (2018) ³²	PPI Histamine 2 receptor antagonist (H2RA)	Dementia	Anti-dementia medication (secondary outcome)	PSSA: SGLT2-I → Gout ± 365 days aSR 0.63 (95%CI 0.47–0.84) PSSA: GLP1-RA → Outcome ± 365 days aSR 0.94 (95%CI 0.78–1.13)
Roughhead (2015) ⁹⁶	Pioglitazone Rosiglitazone	Oedema	Furosemide	PSSA: Rosiglitazone → Furosemide ± 1 year Pooled (Australia and Canada): aSR 1.65 (95%CI 1.58–1.72)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Roughhead (2016) ⁹⁵	PPI	Clostridium difficile infection	Oral vancomycin	PSSA: PPI → Oral vancomycin ± 1 year Pooled estimate: aSR 2.40 (95%CI 1.88–3.05) Pooled estimate (Asia only): aSR 3.16 (95%CI 1.95–5.10) PSSA: Rosiglitazone → Furosemide (Jul 2000 to Dec 2007) aSR = 1.73 (99%CI 1.34–2.24)
Wahab (2014) ¹¹³	Rosiglitazone	Heart failure	Furosemide	
Hachiken (2013) ¹⁰⁹	Low dose aspirin (LDA)	Gastrointestinal (GI) complications	H2RAs PPIs	PSSA: LDA → PPIs ± 365 days Enteric coated LDA: aSR 1.87 (95% CI 1.26–2.83)
Maura (2018) ⁹³	Direct oral anticoagulants (DOACs; excluding edoxaban)	GI events (composite) Nausea Constipation Depression Glaucoma	Gastrointestinal medications (composite) Gastrointestinal medications without acid disorder drugs Antiemetics Drugs for constipation	PSSA: DOAC → Gastrointestinal medications (composite) ± 360 days aSR 0.95 (95%CI 0.92–0.97); n = 24,916 Apixaban → Gastrointestinal medications ± 360 days aSR 1.18 (95%CI 1.10–1.26); n = 34,440 PSSA: DOAC → Gastrointestinal medications (without acid disorder drugs) ± 360 days aSR 1.26 (95%CI 1.24–1.29); n = 37,764
Takada (2014) ⁶⁷	Low dose aspirin (LDA) Enteric coated Buffered	GI complications	H2RAs PPIs	PSSA: DOAC → Antiemetic ± 360 days aSR 1.25 (95%CI 1.22–1.28); n = 27,080 PSSA: DOAC → Drugs for constipation ± 360 days aSR 1.25 (95%CI 1.22–1.27); n = 43,112 DOAC → Antidepressant medication ± 360 days aSR 1.26 (95%CI 1.23–1.30); n = 20,613 DOAC → Glaucoma medication ± 360 days aSR 1.01 (95%CI 0.97–1.05); n = 9473
Yokoyama (2020) ⁸⁴	Oral antiocoagulants	Osteoporosis	Bisphosphonate	PSSA: Warfarin → Bisphosphonate ± 12 months aSR 1.43 (95%CI 1.02–2.03); n = 148
Cardiovascular system	Angiotensin converting enzyme inhibitor (ACEI)	Cough	Antitussive	ACEI → Antitussive (<1 year; adjusted) aOR 1.53 (95%CI 1.17–2.01)
Bowman (1995) ⁷³				(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Fujimoto (2014) ⁵⁰	Statins	Lower urinary tract symptoms (LUTS)	Drugs for storage LUTS	PSSA: Statins → Drugs for storage LUTS ± 365 days All statins: aSR 1.17 (95% CI 1.05–1.30) Pravastatin: aSR 1.27 (95% CI 1.05–1.54) Statins → Solfenacain: aSR 1.47 (95% CI 1.25–1.73) Statins → Oxybutynin: aSR 1.71 (95% CI 1.09–2.72)
Gurwitz (1997) ²³	Antihypertensive medication (see Appendix S1)	Gout	Anti-gout medication (see Appendix S1)	Antihypertensive → Anti-gout medication < 365 days Non-thiazide antihypertensive alone: aRR 1.00 (95% CI 0.65–1.53) Thiazide diuretic alone: aRR 1.99 (95% CI 1.21–3.26) Thiazide diuretic plus non-thiazide antihypertensive: aRR 2.29 (95% CI 1.55–3.37)
Hallas (1996) ⁵²	Beta blockers Cardiovascular drugs (see Appendix S1)	Depression	Antidepressants	Beta-blocker → Antidepressant (study period) aRR 1.09 (95% CI 0.95, 1.26) ACEI → Antidepressant aRR 1.29 (95% CI 1.08, 1.56)
Lindberg & Hallas (1998) ⁹⁸	Cholesterol-lowering medication	Depression	Antidepressants	Calcium channel blockers → Antidepressant aRR 1.31 (95% CI 1.14, 1.51)
Morris (2021) ^{c116}	Dihydropyridine calcium channel blockers (DH-CCBs)	Oedema	Loop diuretic	PSSA: Cholesterol-lowering drug → Antidepressant (study period) All drugs: aSR 0.90 (95% CI 0.68–1.22); n = 184 Simvastatin: aSR 1.59 (1.08–2.45); n = 91
Pouwels (2013) ¹²⁸	ACEI	Urinary tract infection (UTI)	Nitrofurantoin	Among 5458467 DH CCB users (weighted), 185130 individuals (3.4% weighted) were identified with new loop diuretic use. PSSA: ACEI → Nitrofurantoin ± 4 weeks aSR 1.68 (95% CI 1.21–2.36); n = 161
Pouwels (2014) ^{b118}	ACEI	UTI	Nitrofurantoin	ACEI → Nitrofurantoin (<30 days vs <60–90 days) Crude OR = 1.84 (95% CI 1.51–2.25)
Pouwels (2016) ^{e9}	Statin	Infection	Antibiotic	PSSA: Statin → Antibiotic ± 13 months Any antibiotic: aSR 0.86 (95% CI 0.81–0.91)
Pratt (2015) ⁶¹	Amiodarone	Hypothyroidism	Thyroxine	PSSA: Amiodarone → Thyroxine ± 12 months Pooled aSR 2.63 (95% CI 1.47–4.72)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year) ^a
Savage (2020) ⁵	Calcium channel blockers (CCBs) ACEIs or Angiotensin receptor blockers (ARBs) (comparator)	Oedema	Loop diuretic	<i>CCB → Loop diuretic < 90 days</i> Incident CCB users had a higher cumulative incidence of loop diuretic than the comparators [1.4% vs. 0.7% [other antihypertensive comparator] and 0.5% [general comparator], $p < .001$].
				<i>CCB versus other antihypertensive (ACEI or ARB)</i> 1–30 days: aHR 1.68 (95%CI 1.38–2.05) 31–60 days: aHR 2.26 (95%CI 1.76–2.92) 61–90 days: aHR 2.40 (95%CI 1.84–3.13) 91–180 days: aHR 2.24 (95%CI 1.86–2.71) 181–365 days: aHR 1.64 (95%CI 1.38–1.94)
Silwer (2006) ⁹²	Statin	Muscle pain	NSAID	<i>PSSA: Statin → NSAID ± 365 days</i> aSR 0.94 (95%CI 0.85–1.05)
Singh (2021) ⁶⁴	CCBs	Lower extremity oedema	Diuretics	<i>CCB → Diuretic day 8 → day 365</i> Cohort 1: 161 incident diuretic users among 3304 incident CCB users (4.9%, 95%CI 4.2–5.7). Cohort 2: 1586 incident diuretic users among 36462 prevalent CCB users (1.3%, 95%CI 4.1–4.6). Cohort 3: 130 incident diuretic use among 2525 participants with polypharmacy at the day of incident CCB dispensing (5.1, 95%CI 4.3–6.0).
Takada (2014) ¹²⁹	Statins	Sleep disturbance	Hypnotic drugs	<i>PSS: Statin → Hypnotic drugs ± 365 days</i> aSR 1.18 (95%CI 1.11–1.25)
Thieszen (1990) ^{e,112}	Beta-blocker	Depression	Antidepressants	<i>Beta-blocker → Antidepressant < 34 days (concurrent use)</i> Beta-blocker: RR 2.6 (95%CI 2.3–3.0)
Vegter (2013) ¹⁸	ACEI	Cough	Cough medication	<i>PSSA: ACEI → Cough medication ± 6 months</i> 2000–2012: SR 2.0 (95%CI 1.8–2.2)
Vouri (2018) ^{c,6}	DH-CCBs	Lower extremity oedema	Loop diuretic	<i>DH-CCB → Loop diuretic (2014)</i> The potential prescribing cascade was identified in 2.2 million visits (4.6%) using the primary definition of prescribing cascade.
Vouri (2019) ⁷	DH-CCBs	Lower extremity oedema	Loop diuretic	<i>PSSA: DH-CCB → Loop diuretic (2014) ± 360 days</i> aSR 1.87 (95%CI 1.84–1.90)
Vouri (2021) ¹⁰⁵	DH-CCBs	DH-CCB induced oedema	Loop diuretic	<i>PSSA: DH-CCB → Loop diuretic ± 360 days</i> aSR 2.27 (95% CI 1.44–3.58)

(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Vouris (2021) ¹⁰⁴	DH-CCB	DH-CCB induced oedema	Loop diuretic	PSSA: DH-CCB → Loop diuretic ± 360 days Relative to levothyroxine initiators: aSR 1.72 (95%CI 1.66-1.78)
Vouris (2022) ³⁶	Beta-blocker	Oedema	Loop diuretic	Relative to ACEI/ARBs initiators: aSR 1.45 (1.41-1.49)
Yokoyama (2021) ^{d,85}	Amiodarone	Hypothyroidism	Thyroid preparations	PSSA: Beta-blocker → Loop diuretic ± 90 days aSR 1.78 (99%CI 1.72-1.84)
Dermatologicals				
Azoulay (2007) ^{b,45}	Isotretinoin	Depression	Antidepressants	Isotretinoin → Antidepressant (5 month risk and control windows) aRR 2.68 (95%CI 1.10-6.48)
Herson (2003) ⁷²	Isotretinoin Minocycline	Depression	Antidepressants (MAOIs excluded)	Isotretinoin → Antidepressant (study period) aRR 0.97 (95%CI 0.92-1.02)
Sturkenboom (1995) ⁶⁵	Acitretin	Vulvo-vaginal infection	Vulvo-vaginal anti-infective drug	Minocycline → Antidepressant (study period) aRR 0.98 (95%CI 0.95-1.02)
Dyson (2020) ³³	5-α reductase inhibitors (5-ARI)	Depression	Antidepressant	Acitretin → Vulvo-vaginal anti-infective (study period) Pooled Mantel-Haenszel IRR: 3.3 (95%CI 1.1-9.6)
Genito urinary system and sex hormones				
Hagberg (2017) ^{d,110}	5-ARI Alpha blocker (AB)	Depression	Antidepressant (<90 days of depression diagnosis)	PSSA: 5-ARI → Antidepressant ± 365 days Crude SR 0.84 (95% CI 0.80-0.89)
Anti-infectives for systemic use				
Anti-infectives for systemic use				
5-ARIs only: aIRR = 0.94 (95%CI 0.85-1.04)				
5-ARIs + ABs: aIRR = 1.04 (94%CI 0.89-1.21)				
Nested case-control analysis (compared with AB only users)				
5-ARIs only: aOR 0.88 (95%CI 0.78-1.01)				
5-ARIs+ABs: aOR 0.90 (95%CI 0.73-1.10).				

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Corrao (2005) ^{d,49}	Antibacterial drugs for systemic use	Arrhythmia triggered by prolonged QT interval	Antiarrhythmic	PSSA: Antibacterial → Antiarrhythmic (study period) Erythromycin aSR 1.78 (95%CI 1.09, 2.89); n = 73 Ciprofloxacin aSR 1.17 (95%CI 1.02, 1.33); n = 870 Cohort analysis (standardised incidence ratios) Erythromycin: 1.96 (95%CI 1.45–2.59); n = 8956 Clarithromycin: 1.18 (95%CI 1.08–1.29); n = 97900 Rokitamycin: 1.27 (95%CI 1.00–1.66); n = 15247 Ciprofloxacin: 1.25 (95%CI 1.14–1.37); n = 58070 Norfloxacin: 1.17 (95%CI 1.00–1.36); n = 22421 Levofloxacin: 1.33 (95%CI 1.03–1.38); n = 14159 Case–control analysis Erythromycin: OR 1.89 (95%CI 1.33–2.68) Clarithromycin: OR 1.18 (95%CI 1.04–1.34) Ciprofloxacin: OR 1.21 (95%CI 1.05–1.39) Levofloxacin: OR 1.33 (95%CI 1.04–1.70)
Farkas (2021) ²¹	Aromatase inhibitors (AI)	For the treatment of menopausal symptoms Vasomotor symptoms, vaginal dryness, arthralgias, pain	See Appendix S1	Medication use in 12 months before AI: Any new side effect medication: 7436 (40.2%) Opiates 31.5%; SSRIs 16.1%; Gabapentin 7.0% Medication use in the 24 months after AI: Any new side effect medication: 13179 (71.2%) Opiates 55.1%; SSRIs 22.6%; Benzodiazepines 18.4%; Tramadol 17.7%; Gabapentin 14.6%
Gurwitz (1994) ^{a,22}	NSAID	Hypertension	Antihypertensive	NSAID → Antihypertensive (<365 days) OR = 2.01 (95%CI 1.89–2.14)
Avorn (1995) ^{a,19}	Neuroleptics	Extrapyramidal symptoms	APD (excluding amantadine monotherapy)	Any Anti-Parkinson drug (<90 days) Any neuroleptic: aOR 5.4 (95%CI 4.8–6.1) Any anticholinergic anti-Parkinson drug (<90 days) Any neuroleptic: aOR 8.5 (95%CI 4.8–6.1) Dopaninergic agent (<90 days) Any neuroleptic: aOR 2.2 (95%CI 1.9–2.7)
Brandt-Christensen (2007) ³⁷	APD Control 1: Antidiabetics Control 2: unexposed	Depression	Antidepressants	Anti-Parkinson drug → Antidepressant (versus unexposed) APD cohort: RR 2.10 (95%CI 2.04–2.16) Antidiabetic cohort: RR 1.34 (95%CI 1.32–1.36)

(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Dalgard Dunvald (2020) ³⁸	Selective serotonin reuptake inhibitors (SSRI)	Restless leg syndrome (RLS)	Dopamine agonist Quinine	PSSA: SSRI → RLS drug \pm 365 days Any drug: aSR 0.99 (95%CI 0.95–1.02) Dopamine agonist only: aSR 1.21 (95%CI 1.12–1.32); n = 2267
Gau (2010) ^{a,130}	Lithium Carbamazepine Valproate	Hypothyroidism	Thyroxine, leiothyronine or thyroid hormone and hypothyroidism diagnosis (composite)	Likelihood for incident hypothyroidism (study period) Lithium: OR 1.41 (95%CI 1.14–1.74) Carbamazepine: OR 1.37 (95%CI 1.13–1.65) Valproate: OR 1.72 (95%CI 1.40–2.11)
Gill (2005) ²⁶	Acetylcholinesterase inhibitors (AChEI)	Urge urinary incontinence	Urinary anticholinergics	AChEI → Anticholinergic Patients dispensed cholinesterase inhibitors were more likely to receive an anticholinergic medication in follow-up (4.5% vs. 3.1%; p < 0.001)
Hirano (2020) ¹⁰⁰	Anxiolytic Hypnotic Antidepressants Antipsychotics	EPS	Diagnosis of EPS and APD prescription in same month (composite)	PSSA: Psychotropic medication → EPS and APD \pm 12 months Anxiolytic: aSR 2.48 (95%CI 2.16–2.85); n = 992 Hypnotic: aSR 2.28 (95%CI 1.97–2.64); n = 872 Antidepressant: aSR 2.26 (95%CI 1.93–2.66); n = 728 Antipsychotic: aSR 9.24 (95%CI 7.35–11.8); n = 817
Kalisch Ellett (2018) ^{c,114}	Antipsychotics	EPS Hyperprolactinaemia Diabetes mellitus	Anticholinergic Hyperprolactinaemia medications Oral diabetes medications	Concomitant medication use Anticholinergic: n = 51 (0.7%) Hyperprolactinaemia medications: n = 8 (0.1%) Oral diabetes medicines: n = 874 (11.8%)
Kroger (2015) ^{b,117}	AChEI	Urinary incontinence	Drugs for urinary frequency and incontinence	AChEI → Drugs for urinary frequency < 90 days All patients (n = 2700): aHR 1.13 (95%CI 0.97–1.32) Rivastigmine patients (n = 1853): aHR 1.13 (95%CI 0.95–1.34) Galantamine patients (n = 1043): aHR 1.10 (95%CI 0.81–1.50)
Lai (2013) ⁷⁹	Antiepileptic drugs (AEDs)	Hypothyroidism	Levothyroxine	PSSA: AEDs → Levothyroxine \pm 12 months Any AED: aSR 1.13 (99%CI 1.09–1.18) Carbamazepine: aSR 1.21 (99%CI 1.08–1.34) Phenobarbital: aSR 1.25 (99%CI 1.15–1.36) Phenytoin: aSR 1.75 (99%CI 1.58–1.94) Valproate: aSR 1.34 (99%CI 1.20–1.49) Oxcarbazepine: aSR 1.22 (99%CI 1.03–1.46)
Lampela (2016) ⁴⁴	AChEI or Memantine	Urinary incontinence	Urinary anticholinergics	AChEI → Urinary anticholinergics (versus memantine users) <6 months: aHR 1.47 (95%CI 1.17–1.86) <12 months: aHR 1.41 (95%CI 1.17–1.69)
Marras (2016) ²⁷	Lithium Valproic acid Antidepressant	Drug induced tremor diagnosed as Parkinson's Disease (PD)	Anti-Parkinson drug or PD diagnosis (see Appendix S1)	Start of dopaminergic drug (no previous antidepressant): aHR (95%CI 1.06–3.30) Start of anti-Parkinson drug or PD diagnosis (no previous antidepressant use) Lithium (versus antidepressant): aHR 1.68 (95%CI 1.13–2.48)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Masurkar ²⁴ (2021) ²⁴	AChEI	Overtactive bladder	Urinary anticholinergic	AChEI → Anticholinergic cascade <6months Rivastigmine: aHR = 1.0 Donepezil: aHR = 1.55 (95%CI 1.31-1.83) Galantamine: aHR = 1.17 (95%CI 0.87-1.58)
Movig (2002) ⁴¹	SSRI	Urinary incontinence	Spasmolytic agent or 30 or more units of incontinence wear	SSRI → Spasmolytic agent/incontinence wear <3month During SSRI (versus before SSRI): IDR 1.57 (95%CI 1.38-1.79) During SSRI (versus after SSRI): IDR 2.03 (95%CI 1.76-2.34) During SSRI (versus before and after SSRI): IDR 1.75 (95%CI 1.56-1.97) <i>Risk for incontinence during exposed period (versus non-exposed)</i> aRR 1.61, 95%CI 1.42-1.82
Narayan (2019) ²⁵	AChEI or Memantine	Several ADRs examined relating to anticholinergic medication use	Anticholinergics (see Appendix S1)	<i>Anti-dementia drug → Marker medication ± 180 days</i> Exposed to at least one anticholinergic ±180days: n = 1439 Exposed to at least one anticholinergic after anti-dementia drug: n = 416
Onder (2014) ^{c,46}	Anti-Parkinson drugs and antipsychotics (concomitant use)	Parkinsonism (side effect of antipsychotics); Behavioural disorders (side effect of anti-Parkinson drugs)	Anti-Parkinson drugs and antipsychotics (concomitant use)	<i>Prevalence of concomitant use of anti-Parkinson and antipsychotic medication (2011)</i> Total population: n = 25 949 (0.2%) 65-74 years: n = 10 200 (0.2%) 75-84 years: n = 10 625 (0.2%) ≥85 years: n = 5124 (0.3%) PSSA: Benzodiazepines → Anti-dementia drugs ± 3years aSR 2.19 (95%CI 1.92-2.49); n = 1285
Park (2018) ⁹⁴	Benzodiazepines	Dementia	Anti-dementia drugs	Flunarizine → Antidepressant <30days
Petri (1988) ⁵⁵	Flunarizine	Depression	Antidepressant	Number of antidepressant starts during or within 30days after flunarizine use was 5 out of a total of 34 histories
Petri (1990) ⁵⁷	Flunarizine	Depression or Parkinsonism	Antidepressant or Anti-Parkinson drug	Flunarizine → Antidepressant (study period) Incidence Rate = 1.342 (95%CI 1.00-1.80) Flunarizine → Anti-Parkinson drug
Pratt (2013) ⁶⁰	Antipsychotics	Acute hyperglycaemia	Insulin	In a subset of 777 flunarizine recipients there were 10 participants who received anti-Parkinson drugs PSSA: Olanzapine → Insulin ± 12 months USA Public: aSR 1.14 (95%CI 1.1-1.17) Sweden: aSR 1.53 (95%CI 1.13-2.06) Risperidone → Insulin ± 12 months USA Public: aSR 1.09 (95%CI 1.07-1.12)
Read (2021) ²⁸	Gabapentinoid	Oedema	Diuretic	Gabapentinoid → Diuretic < 90days (versus non-users) aHR 1.44 (95%CI 1.23-1.70).

(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Rochon (2005) ²⁹	Antipsychotic	Parkinsonism	Anti-Parkinson drug or Parkinson diagnosis (composite)	Antipsychotic → Anti-Parkinson drug/diagnosis <1 year (versus atypical antipsychotic) Typical antipsychotics: adjusted HR 1.30 (95% CI 1.04–1.58) No therapy: aHR 0.40 (95% CI 0.29–0.43)
Takada (2016) ⁴⁶	Benzodiazepine	Dementia	Anti-dementia drug	PSSA: Benzodiazepine → Anti-dementia drug ± 12 months Olanzapine ±360 days: aSR 1.23 (95% CI 1.11–1.37)
Takeuchi (2015) ⁴³	Atypical antipsychotics	Hyperlipidemia	Anti-hyperlipidemic drugs	PSSA: Atypical antipsychotics → Anti-hyperlipidemic drugs Fully-adjusted RR = 1.19 (95% CI 0.52–2.74)
Thacker (2006) ¹⁰⁷	AChEI	Drug-induced airways complications	Antibacterial and oral corticosteroid	AChEI → Marker drug ± 1 year Loperamide/Oral rehydration: aSR 1.42 (95% CI 1.14–1.77); <i>n</i> = 348
Venalainen (2017) ⁷⁰	AChEI	Nausea Dyspepsia Diarrhoea Urinary incontinence Seizures Anxiety Insomnia Depression	Antiemetics PPIs/H2RAs Loperamide/Oral rehydration sachets Oxybutynin Anxiolytics Anticonvulsants Hypnotics and sedatives Antidepressants	Anxiolytics: aSR 1.16 (95% CI 1.01–1.34); <i>n</i> = 807 Hypnotics and sedatives: aSR 1.19 (95% CI 1.05–1.36); <i>n</i> = 963 Antiemetics: aSR 1.18 (95% CI 1.05–1.32); <i>n</i> = 1202 Anticonvulsants: aSR 1.26 (95% CI 1.03–1.5); <i>n</i> = 389 PPI/H2RAs: aSR 0.87 (95% CI 0.77–0.98), <i>n</i> = 1079 Antidepressant: aSR 0.77 (95% CI 0.70–0.85), <i>n</i> = 1698 Oxybutynin: aSR 1.04 (95% CI 0.81–1.34), <i>n</i> = 261
Vouri (2020) ^{c,131}	AChEI or Memantine	Rhinorrhea	Rhinorrhea medications (see Appendix S1)	AChEI/Memantine → Rhinorrhea medications (concomitant use) AChEI users were more likely to use a rhinorrhea medication compared to non-AChEI users, OR 7.16 (95% CI 2.25–22.73); adjusted OR = 4.7 (95% CI 1.53–14.43)
Wang (2021) ⁸⁰	Varenicline	Neuropsychiatric adverse events: Depression Anxiety Sleep disorders	Antidepressant Anxiolytics Hypnotics and sedatives (composite outcome)	PSSA: Varenicline → Any NPAE drug ± 365 days aSR 1.00 (95% CI 0.89–1.13)
Wang (2021) ⁴²	Varenicline (Nicotine replacement therapy [NRT] as comparator)	Neuropsychiatric adverse events: Depression Anxiety Insomnia	Antidepressants Antidepressants in combination with psycholeptics Anxiolytics Hypnotics and sedatives (composite outcome)	PSSA: Varenicline → Hypnotics and sedatives ± 365 days Sleep disorder drug: aSR = 1.25 (95% CI 1.05–1.48) General population with psychiatric disorders <24 weeks Any NPAE medication: adjusted OR 0.82 (95% CI 0.68 to 0.99) General population without psychiatric disorders <24 weeks Any NPAE medication: adjusted OR 0.85, (95% CI 0.72 to 1.00) COPD population with psychiatric disorders <24 weeks Any NPAE medication: adjusted OR 0.97 (95% CI 0.66 to 1.44) COPD population without psychiatric disorders <24 weeks Any NPAE medication: adjusted OR 0.81 (95% CI 0.54 to 1.20)
Yokoyama (2020) ^{b,86}	Antipsychotics	Osteoporosis	Bisphosphonate	PSSA: Antipsychotic → Bisphosphonate No association identified.

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Respiratory system Fox (2022) ³⁴	Montelukast	Neuropsychiatric adverse events (NPAE)	Antidepressants Benzodiazepines Hypnotics Antipsychotics Mood stabilisers Buspirone (composite outcome)	PSSA: Montelukast → Any NPAE medication ± 14–365 days SR 0.84 (95%CI 0.80–0.89)
Henriksen (2017) ³⁹	Inhaled corticosteroids	Oral candidiasis	Systemic or topical antifungal	PSSA: Inhaled corticosteroid → Topical antifungal ± 12 months Crude SR 2.89 (95%CI 2.80–2.97)
Petri (1991) ⁵⁶	Inhaled corticosteroids	Oral candidiasis	Topical antifungal	PSSA: Inhaled corticosteroid → Systemic antifungal ± 12 months Crude SR 1.50 (95%CI 1.46–1.54)
Van Boven (2013) ⁷¹	Inhaled corticosteroids	Oral candidiasis	Topical antifungal	Inhaled corticosteroids → Topical antifungal < 90 days Crude OR = 1.66 (n = 21)
Winkel (2018) ⁴⁰	Montelukast	Depression	Antidepressant (excluding bupropion)	PSSA: Montelukast → Antidepressant ± 1 year Crude SR 1.19 (95%CI 1.11–1.28)
Sensory organs Roughhead (2012) ⁹⁷	Timolol Latanoprost Bimatoprost Pilocarpine Brimonidine	Exacerbation of airways disease Exacerbation of depression	Inhaled beta-agonists Inhaled corticosteroids Oral corticosteroids SSRI	PSSA: Glaucoma → marker medications ± 1 year Timolol → Inhaled beta agonist: aSR 1.48 (95%CI 1.22–1.78); n = 786 Timolol → Inhaled corticosteroid: aSR 1.43 (95%CI 1.13–1.81); n = 494 Latanoprost → Inhaled beta agonist: aSR 1.24 (95%CI 1.11–1.38); n = 2251 Latanoprost → Oral corticosteroid: aSR 1.14 (95%CI 1.00–1.29); n = 1671 Timolol → Antidepressant: aSR 1.24 (95%CI 1.07–1.43); n = 1253 Timolol → SSRI: aSR 1.30 (95%CI 1.08–1.56); n = 791 Latanoprost → Antidepressant: aSR 1.16 (95%CI 1.03–1.31); n = 1871 Latanoprost → SSRI: aSR 1.20 (95%CI 1.03–1.39); n = 1155
Multiple medication groups examined Brandt-Christensen (2006) ⁸²	Antidepressant Lithium Antidiabetic	Parkinsonism	APD (see Appendix S1 for exclusions)	Index drug → Anti-Parkinson drug (versus unexposed) Antidepressant: RR 1.79 (95%CI 1.72–1.86) Lithium: RR 1.88 (95%CI 1.60–2.20) Antidiabetic: RR 0.80 (95%CI 0.74–0.86)

(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Bytzer & Hallas (2000) ⁸¹	Predefined list of 32 index medications (see Appendix S1)	Dyspepsia or nausea	Cisapride or Metoclopramide	PSSA: Index medication → Cisapride < 100 days NSAIDs: aSR = 1.33 (95%CI 1.02–1.77); n = 211 Methylxanthines: aSR = 2.36 (1.00–8.44); n = 18 PSSA: Index medication → Metoclopramide < 100 days Insulin aSR 2.91 (95%CI 1.40–8.11); n = 28 Opioids: aSR 2.84 (95%CI 2.48–3.28); n = 1017 Potassium supplement: 1.42 (95%CI 1.15–1.79); n = 324 Digoxin: 2.87 (95%CI 2.01–4.35); n = 138 Nitrates: 1.74 (95%CI 1.16–2.77); n = 88 Loop diuretics: 1.50 (95%CI 1.23–1.85); n = 383 ACEIs: 2.27 (95%CI 1.46–3.85); n = 77 Oral corticosteroids: 1.33 (95%CI 1.11–1.60); n = 458 Antibiotics: 1.40 (95%CI 1.24–1.60); n = 974 Penicillins: 1.38 (95%CI 1.21–1.59); n = 868 Macrolides: 1.58 (95%CI 1.31–1.94); n = 414 NSAIDs: 1.48 (95%CI 1.28–1.74); n = 676 Asthma drugs: 1.42 (95%CI 1.14–1.79); n = 307 Methylxanthines: 2.03 (95%CI 1.25–3.65); n = 63 PSSA: Index medication → Prochlorperazine ± 12 months Cardiac therapy: aSR = 1.14 (95%CI 1.06–1.22); n = 3017 Nitrates: aSR = 1.11 (95%CI 1.03–1.21); n = 2224 Isosorbide mononitrate: aSR = 1.21 (95%CI 1.07–1.38); n = 918 Diuretic: aSR = 1.07 (95%CI 1.01–1.14); n = 3845 Beta-blocker: aSR = 1.13 (95%CI 1.05–1.21); n = 3156 CCBs: aSR = 1.22 (95%CI 1.16–1.36); n = 2896 ACE inhibitors: aSR = 1.22 (95%CI 1.14–1.31); n = 3162 AR2B: aSR = 1.20 (95%CI 1.11–1.30); n = 2577 Statins: aSR = 1.50 (95%CI 1.40–1.61); n = 3411 NSAIDs: aSR = 1.22 (95%CI 1.16–1.36); n = 2896 ACE inhibitors: aSR = 1.22 (95%CI 1.14–1.31); n = 3162 AR2B: aSR = 1.20 (95%CI 1.11–1.30); n = 2577 Statins: aSR = 1.50 (95%CI 1.40–1.61); n = 3411 NSAIDs: aSR = 1.37 (95%CI 1.27–1.47); n = 3079 Opioids: aSR = 1.24 (95%CI 1.17–1.31); n = 5266 Sedatives: aSR = 1.18 (95%CI 1.11–1.26); n = 3470 Antidepressant → Ulcer drugs (compared with TCA only) SSRI: IRR 1.2 (95%CI 0.5–2.8); n = 1181 SSRI + NSAID: IRR 12.4 (95%CI 3.2–48.0); n = 86
de Jong (2003) ⁴⁸	Medicines commonly associated with dizziness (identified (see Appendix S1))	Dizziness	Prochlorperazine	H2RAs PPIs Prostaglandins
de Jong (2003) ¹⁰⁸	Antidepressant with or without NSAID			GI adverse effects

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Garrison (2012) ⁵¹	Statin Diuretic Inhaled long-acting beta-agonists (LABA)	Nocturnal leg cramps	Quinine	PSSA: Index drug → Quinine ± 1 year All statins: aSR 1.16 (95%CI 1.04–1.29); n = 1326 All LABAs: aSR 2.42 (95%CI 2.02–2.89); n = 576 LABA alone: aSR 2.17 (95%CI 1.56–3.02); n = 137 LABA-corticosteroid: aSR 2.55 (95%CI 2.06–3.12); n = 439 All diuretics: aSR 1.47 (95%CI 1.33–1.63); n = 1590 Loop diuretic: aSR 1.20 (95%CI 1.00–1.44); n = 447 Thiazide diuretic: aSR 1.48 (95%CI 1.29–1.68); n = 977 Potassium-sparing diuretic: aSR 2.12 (95%CI 1.61–2.78); n = 206
Hallas & Bytzer (1998) ⁸⁹	Predefined list of 33 medications (see Appendix S1)	Dyspepsia	Ulcer drug prescription	PSSA: Index drug → Ulcer drug prescription ± 100 days NSAIDs: aSR 1.80 (95%CI 1.64–1.99) CCBs: aSR 1.40 (95%CI 1.18–1.67) Oral corticosteroids: aSR 1.15 (95%CI 1.02–1.30) ACEIs: aSR 1.38 (1.12–1.73) Methylxanthines: aSR 1.49 (1.05–2.19)
Hashimoto (2015) ⁵³	Medicines that cause storage symptoms; Medicines that cause voiding symptoms	LUTS	Medications for (LUTS)	PSSA: Index drug → Medications for LUTS ± 12 months Oxycodone: aSR 1.20 (95%CI 1.03–1.41) Morphine: aSR: 1.29 (95%CI 1.14–1.45) Donepezil: aSR: 1.98 (95%CI 1.57–2.50) Intestinal lavage solution: aSR 1.86 (95%CI 1.65–2.10) Cyclophosphamide: aSR 1.52 (95%CI 1.14–2.04) Levodopa/benserazide: aSR 1.82 (95%CI 1.18–2.81) Amanitadine: aSR 1.53 (95%CI 1.12–2.09) Paroxetine: aSR 1.77 (95%CI 1.33–2.36) Milnacipran: aSR 2.10 (95%CI 1.28–3.45) Diazepam: aSR 1.73 (95%CI 1.46–2.06) Risperidone: aSR 1.55 (95%CI 1.34–1.79) Levomeprazine: aSR 2.20 (95%CI 1.34–1.79) Sulpiride: aSR 1.32 (95%CI 1.01–1.72) Cimetidine: aSR 1.99 (95%CI 1.24–3.20) Scopolamine butylbromide: aSR 1.72 (95%CI 1.55–1.92) Tiotropium bromide: aSR 1.75 (95%CI 1.42–2.16) Cibenzoline: sSR 2.97 (95%CI 1.92–4.59) Amezinium metilsulfate: aSR 1.89 (95%CI 1.10–3.26)
Huh (2019) ³¹	Metoclopramide or levosulpiride	Drug induced Parkinsonism	Levodopa	PSSA: Metoclopramide → Levodopa < 90 days aOR 2.94 (95%CI 2.35, 3.67) PSSA: Levosulpiride → Levodopa < 90 days aOR 3.30 (95%CI 2.52, 4.32)

(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Kalisch Ellett (2014) ⁷⁴	See Appendix S1	Urinary incontinence	Oxybutynin	PSSA: Index medication → Oxybutynin ± 12 months Prazosin (women only): aSR 1.84 (95%CI 1.29–2.63); n = 135 Low-ceiling diuretics, excluding thiazides: aSR 1.22 (95%CI 1.06–1.41); n = 750 CCBs: aSR 1.45 (95%CI 1.33–1.57); n = 2230 ACEIs: aSR 1.28 (95%CI 1.19–1.39); n = 2616 ACEIs + diuretic: aSR 1.35 (1.15–1.58); n = 620 ARBs: aSR 1.42 (1.30–1.55); n = 2040 ARB+ diuretic: aSR 1.32 (1.16–1.49); n = 99 HRT: aSR 1.54 (95%CI 1.42–1.67); n = 2446 Antipsychotics: aSR 0.83 (95%CI 0.78–0.89); n = 2121 Hypnotic sedatives: aSR 1.10 (95%CI 1.03–1.18); n = 3326 <i>Index medication → Anti-Parkinson drug/diagnosis <1 year</i> Levosulpiride: OR 4.3 (95%CI 3.5–5.3); n = 595 Mosapride: OR 2.1 (95%CI 1.7–2.6); n = 430 Domperidone: OR 2.1 (95%CI 1.6–2.8); n = 247 Metoclopramide: OR 2.7 (95%CI 1.8–4.1); n = 121 Itopride: OR 1.6 (95%CI 1.2–2.2); n = 232 Clebopride: OR 12.8 (95%CI 2.8–57.0); n = 19 Combined propulsive use: OR 3.9 (95%CI 2.8–5.5); n = 219 Typical antipsychotic: OR 6.4 (95%CI 1.4–28.2); n = 17 Atypical antipsychotic: OR 2.4 (95%CI 1.2–4.9); n = 56 Risperidone: OR 13.5 (95%CI 1.8–102.1); n = 23 Flunarizine: OR 5.0 (95%CI 2.7–9.0); n = 86 <i>PSSA: Statin → Antibiotic ± 365 days</i> aSR 1.40 (95%CI 1.34–1.47); n = 7726 <i>PSSA: Statin → Antidiabetic ± 365 days</i> aSR 1.09 (95%CI 1.04–1.15); n = 6794 <i>PSSA: Antidiabetic → Antibiotic ± 365 days</i> aSR 1.24 (95%CI 1.15–1.33); n = 2828
Kim (2019) ^{a,119}	Propulsives Antipsychotics Antivertigo agent (see Appendix S1)	Drug induced Parkinsonism	APD or Parkinson diagnosis (composite) (see Appendix S1)	
Ko (2019) ⁷⁶	Statins Statins	Skin and soft tissue infection New onset diabetes mellitus	Dicloxacillin/Flucloxacillin Antidiabetic	

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Nishtala & Chyou (2017) ⁵⁴	Amiodarone Lithium Frusemide Fluticasone Simvastatin	Hypothyroidism Hyperthyroidism Hypokalaemia Oral candidiasis Muscle cramps	Thyroxine Carbimazole Potassium Nystatin Quinine sulphate	PSSA: Amiodarone → Thyroxine ± 360 days aSR 3.57 (95% CI 3.17–4.02) Lithium → Thyroxine ± 360 days aSR 3.43 (95% CI 2.55–4.70) Amiodarone → Carbimazole ± 360 days aSR 8.81 (95% CI 5.86–13.77) Simvastatin → Quinine sulphate ± 360 days aSR 1.69 (95% CI 1.61–1.77) Fluticasone → Nystatin ± 360 days aSR 2.34 (95% CI 2.19–2.50) Frusemide → Potassium ± 360 days aSR 2.94 (95% CI 2.83–3.05)
Pouwels (2013) ¹³²	SSRI with or without NSAID	Peptic ulcer	Peptic ulcer drug treatment	PSSA: SSRI +/- NSAID → Peptic ulcer treatment ± 4 weeks SSRI: aSR 0.83 (95% CI 0.65–1.06) NSAID: aSR 2.50 (95% CI 2.27–2.76) SSRI + NSAID: aSR 1.48 (95% CI 0.90–2.49)
Rasmussen (2015) ⁶²	Antithrombotic drugs Cardiovascular drugs (see Appendix S1)	Erectile dysfunction	5-phosphodiesterase inhibitor	PSSA: Cardiovascular drugs → 5-phosphodiesterase inhibitor ± 6 months Thiazides: aSR 1.28 (95% CI 1.20, 1.38); NNTH 370 (95% CI 300, 500); n = 3118 β-blockers: aSR 1.18 (95% CI 1.09, 1.28); NNTH 680 (95% CI 480, 1200); n = 2511 CCBs: aSR 1.29 (95% CI 1.21, 1.38); NNTH 330 (95% CI 270, 440); n = 3379 ACEIs: aSR 1.29 (95% CI 1.21, 1.37); NNTH 350 (95% CI 290, 440); n = 4182 ARBs: aSR 1.16 (95% CI 1.06, 1.26); NNTH 540 (95% CI 360, 1200); n = 2082
Singh (2021) ⁶³	Antipsychotic or Metoclopramide	Parkinsonism	Anti-Parkinson drug	Anti-psychotic/metoclopramide → Anti-Parkinson drug < day 8–365 Cohort 1: 36 (0.8%) incident anti-Parkinson drug users among 4534 incident antipsychotic/metoclopramide users Cohort 2: 20 (0.5%) incident users of anti-Parkinsonian drugs among 3485 antipsychotic/metoclopramide users AChEIs → Urinary medications < 6 months 60 cases of prescribing cascade were identified. Extending to 365 days resulted in 52 additional cases. Metoclopramide → Anti-Parkinson drug < 6 months 11 cases of the prescribing cascade were identified. Extending to 365 days resulted in 5 additional cases. CCB → Diuretic < 6 months 289 cases of prescribing cascade were identified. Extending to 365 days resulted in 369 cases.
Treneman (2021) ³⁰	AChEIs Metoclopramide CCBs	Urinary incontinence Parkinsonism Pedal oedema	Urinary medications Anti-Parkinson drug Diuretic	(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Tsiropoulos (2009) ⁶⁸	AEDs	Exploratory analysis	Any other medication presented in the same period	PSSA: All AEDs → Marker medication Propulsives ±183 days: aSR 1.31 (95%CI 1.11-1.56); n = 571 Laxatives ±183 days: aSR 1.57 (95%CI 1.29-1.92); n = 432 Topical corticosteroids ±183 days: aSR 1.32 (95%CI 1.16-1.52); n = 900
				PSSA: Carbamazepine → Marker medication Propulsives ±183 days: aSR 1.57 (95%CI 1.14-2.19); n = 163 Laxatives ±183 days: aSR 1.61 (95%CI 1.01-2.59); n = 82 Topical corticosteroids ±183 days: aSR 1.48 (95%CI 1.17-1.87); n = 305
King (2020) ⁷⁵	654 different medications examined	New onset heart failure	Furosemide	PSSA: Oxcarbazepine → Marker medication Propulsives ±183 days: aSR 2.54 (95%CI 1.71-3.85); n = 119 Laxatives ±183 days: aSR 3.74 (95%CI 2.31-6.29); n = 103 Topical corticosteroids ±183 days: aSR 1.40 (95%CI 1.08-1.83); n = 245 Phenobarbital → Marker medication Bone disease treatment ±548 days: aSR 1.98 (95%CI 1.03-3.92); n = 18

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year) ^a
Wahab (2016) ¹⁰⁶	691 different medications examined	Heart failure	Furosemide	PSSA: Index medication → Furosemide ± 1 year Teriparatide: aSR 5.02 (95% CI 1.07–23.7); n = 10 Lodoxamide: aSR 2.50 (95% CI; 1.06–5.91); n = 27 Famotidine: aSR 1.69 (95% CI 1.38–2.08); n = 423 Latanoprost: aSR 1.48 (95% CI 1.38–1.59); n = 3107 Pilocarpine: aSR 1.43 (95% CI 1.16–1.77); n = 632 Brinzolamide: aSR 1.37 (95% CI 1.16–1.62); n = 564 Betahistidine: aSR 1.31 (95% CI 1.07–1.62); n = 359 Ranitidine: aSR 1.24 (95% CI 1.17–1.31); n = 5554 Paracetamol: aSR 1.06 (95% CI 1.04–1.09); n = 24210
Chen (2021) ⁸⁷	Confirmatory analysis Amiodarone Exploratory analysis ACEIs Statins Buffered LDA Enteric-coated LDA DH-CCBs	Hypothyroidism Gout Cough UTI Storage LUTS Depression Sleep disturbances Hepatotoxicity Muscle pain Skin and soft tissue infection Infection in those with type-2 diabetes GI complications Oedema		Confirmatory PSSA ± 1 year Amiodarone → Thyroxine: aSR 3.77 (95% CI 3.43–4.14); n = 2667 Amiodarone → Allopurinol: aSR 0.83 (95% CI 0.76–0.90); n = 2071 Exploratory PSSA ± 1 year ACEIs → Antitussive: aSR 1.33 (95% CI 1.31–1.34); n = 141924 Statins → Drugs for urinary frequency: aSR 1.17 (95% CI 1.16–1.19); n = 107422 Statins → Antidepressants: aSR 1.19 (95% CI 1.18–1.21); n = 117443 Statins → Hypnotics: aSR 1.10 (95% CI 1.09–1.12); n = 124061 Statins → Ursodeoxycholic acid: aSR 1.26 (95% CI 1.21–1.31); n = 11231 Statins → NSAIDs: aSR 1.02 (95% CI 1.02–1.03); n = 430774 Statins → Dicloxacillin/Flucloxacillin: aSR 1.18 (95% CI 1.15–1.22); n = 23068 Statins → Antibiotic treatment (those with type 2 diabetes): aSR 1.38 (95% CI 1.36–1.39); n = 150016 DH-CCBs → Loop diuretic: aSR 1.46 (95% CI 1.45–1.48); n = 139375

(Continues)

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Lai (2014) ⁷⁸	Sulpiride Non-sulpiride antipsychotics	EPS Diabetes Hyperprolactinaemia Cardiac arrhythmias	Confirmatory analyses: Anticholinergics Oral hypoglycaemics Prolactine inhibitors Class 1B antiarrhythmics Exploratory analyses: all medications prescribed after the index date	Confirmatory PSSA analyses ± 12 months Sulpiride → Anticholinergics: aSR 1.73 (95%CI 1.46–2.06); n = 568 Haloperidol → Anticholinergics: aSR 1.99 (95%CI 1.68–2.35); n = 611 Risperidone → Anticholinergics: aSR 1.21 (95%CI 1.04–1.41); n = 702 Olanzapine → Anticholinergics: aSR 0.73 (95%CI 0.58–0.93); n = 281 Amisulpiride → Anticholinergics: aSR 0.54 (95%CI 0.40–0.73); n = 188 Sulpiride → Prolactine inhibitors: aSR 12.0 (95%CI 1.59–91.2); n = 16 Amisulpiride → Prolactine inhibitors: aSR 8.05 (95%CI 1.00–65.4); n = 8 Haloperidol → Class 1b antiarrhythmics: sSR 2.81 (95%CI 1.03–7.66); n = 21 Exploratory PSSA analyses: Sulpiride → Marker medication ± 12 months Stomatological preparations: aSR 1.86 (95%CI 1.13–3.07); n = 71 Corticosteroids for local oral treatment: aSR 1.71 (95%CI 1.00–2.91); n = 59 Beta blockers, any: aSR 1.42 (95%CI 1.12–1.71); n = 371 Beta blockers, non-selective: aSR 1.61 (95%CI 1.28–2.03); n = 304 Dermatological preparations, corticosteroids: aSR 2.18 (95%CI 1.21–3.92); n = 57 Corticosteroids weak, other combinations: aSR 2.15 (95%CI 1.08–4.28); n = 42 Quinolones: SR 1.50 (95%CI 1.00–2.24); n = 101 Fluroquinolones: aSR 1.81 (95%CI 1.03–3.17); n = 55 Anti-inflammatory preparations, non-steroidal for topical use: aSR 1.36 (95%CI 1.01–1.84); n = 173

TABLE 1 (Continued)

Primary author (year)	Initial medication(s)	Suspected ADR	New medication(s)	Quantitative association (primary analysis or association at 1 year)
Hallas (2018) ¹⁰¹	186 758 associations tested in the main analysis; 30 best signals reported	Exploratory analysis	30 strongest signals reported	PSSA: Index → Marker medication ± 12 months Opioids → Drugs for constipation (crude SR 2.34, 95%CI 2.31–2.38); n = 84 020 High ceiling diuretics → Potassium SR 3.31 (95%CI 3.24–3.38); n = 48 539 Thiazide → Potassium SR 3.46 (95%CI 3.39–3.54); n = 45 175 Opioids → Propulsives SR 2.14 (95%CI 2.10–2.17); n = 62 139 NSAIDs → Anti-ulcer drugs SR 1.71 (95%CI 1.67–1.74); n = 49 646 Antithrombotic → Anti-ulcer drugs SR 1.41 (95%CI 1.39–1.44); n = 54 841 Cough suppressants → Drugs for constipation SR 1.95 (95%CI 1.90–2.00); n = 26 0015 Corticosteroids, systemic use → Drugs affecting bone structure and mineralisation SR 3.40 (95%CI 3.27–3.54); n = 13 023
Helfritzsch (2018) ¹⁰²	Non-vitamin K oral anticoagulants (NOAC)	Exploratory analysis	20 strongest signals reported	PSSA: NOAC → Marker drug ± 6 months Benzodiazepines, hypnotic: cSR 8.28 (95%CI 6.01–12.05); NNTH 193 Osmotic laxatives: cSR 1.35 (95%CI 1.25–1.46); NNTH 133 Benzodiazepines, sedative: cSR 1.99 (95%CI 1.74–2.30); NNTH 174 Corticosteroids, anal use: cSR 2.03 (95%CI 1.76–2.35); NNTH 176 SSRI: cSR 1.57 (95%CI 1.37–1.77); NNTH 202 Other antidepressant: cSR 1.59 (95%CI 1.41–1.80); NNTH 207 PPI: cSR 1.19 (95%CI 1.11–1.28); NNTH 209 Phenylpiridine opioids: cSR 2.12 (95%CI 1.81–2.54); NNTH 215 Propulsives: cSR 1.51 (95%CI 1.35–1.71); NNTH 216 Iron bivalent, oral: cSR 1.62 (95%CI 1.42–1.86); NNTH 238 Contact laxatives: cSR 1.29 (95%CI 1.17–1.43); NNTH 253

Abbreviations: aHR, adjusted hazard ratio; aIRR, adjusted incidence rate ratio; aOR, adjusted odds ratio; aSR, adjusted sequence ratio; cSR, crude sequence ratio; HR, hazard ratio; IDR, incidence density ratio; IR, incidence rate ratio; NNTH, number needed to harm; PSSA, prescription sequence symmetry analysis.

^aCase-control study.

^bCase-crossover study.

^cCross-sectional study.

^dIncludes case-control study.

^eIncludes cross-sectional study.

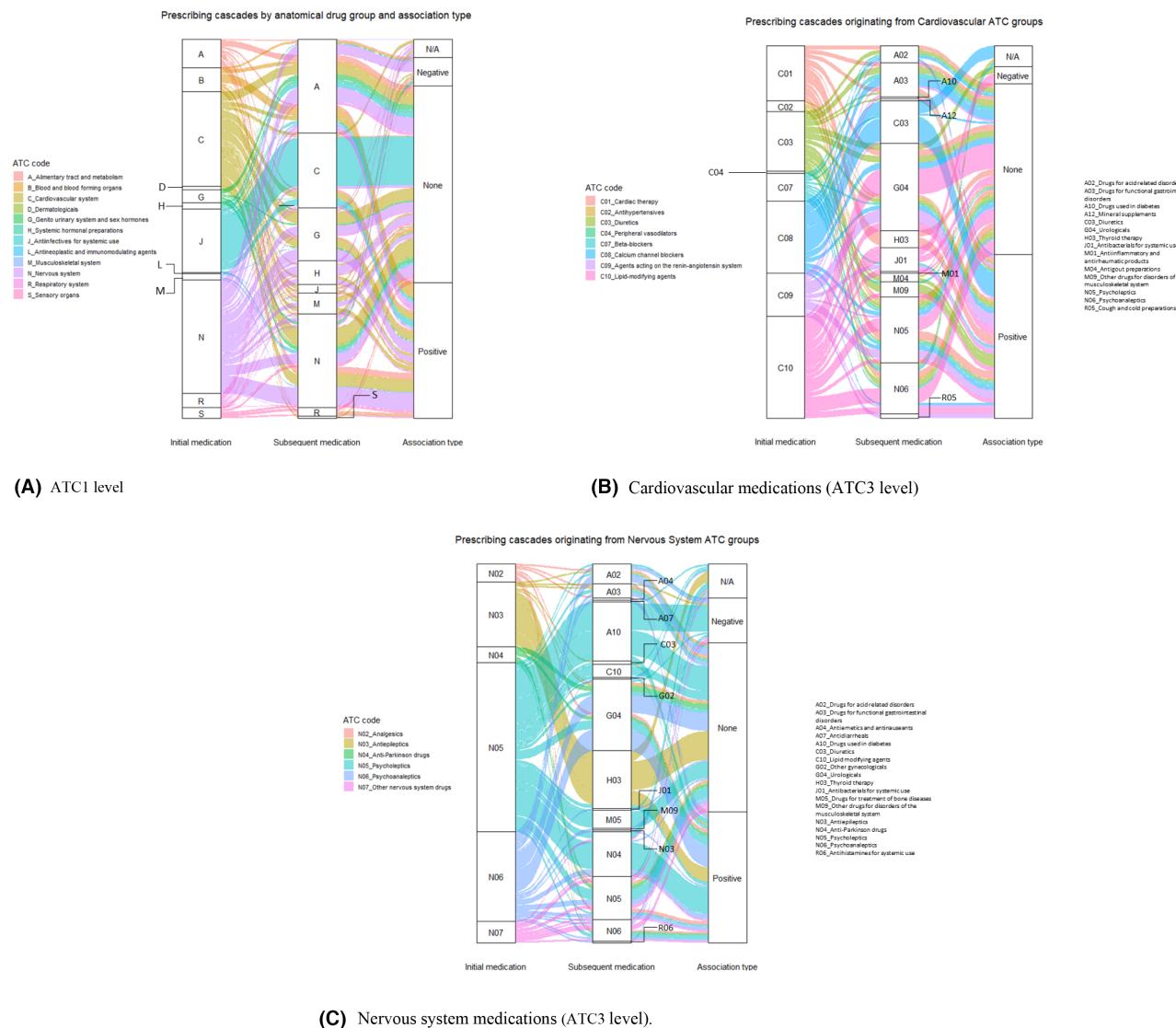


FIGURE 2 Prescribing cascades examined in non-exploratory studies ($n = 94$) stratified by ATC classification. These alluvial plots represent initial (column 1) and subsequent (column 2) medication pairs examined and the primary quantitative association identified (column 3). The height of the strata in columns 1 and 2 is proportional to the number of instances that the relevant medication has been examined across included studies. The height of the strata in column 3 is proportional to the number of identified quantitative associations that belong to each association type. The width of the linkage between column 1 and column 2 is proportional to the number of instances that the unique medication pair has been examined across included studies. The width of the linkage between column 2 and column 3 is proportional to the number of tested medication pairs that result in a prescribing cascade (positive association), do not result in a prescribing cascade (none), indicate a lower likelihood of a prescribing cascade (negative association), or where no association could be examined due to study reporting (N/A: non-applicable); (A) ATC1 level; (B) Cardiovascular medications (ATC3 level); (C) Nervous system medications (ATC3 level).

Figure 2.^{68,75,78,87,101,102,106} Initial medication Anatomical Therapeutic Classification (ATC) codes were not reported for 66 studies and were assigned by our research team.

3.5 | Suspected adverse reaction(s)

Throughout the included studies, suspected ADRs were presumed to have occurred based on the initiation of the second medication

as a treatment. In one study examining the CCB → loop diuretic prescribing cascade, an additional medical chart review was also conducted.¹⁰⁵

The suspected ADRs, symptoms or new diagnoses explored were broad-ranging (see Table 2) most commonly depression ($n = 13$)^{33,37,40,45,52,55,57,72,93,97,98,110,112}; peripheral oedema ($n = 11$)^{5-7,28,30,36,64,96,103-105,116}; urinary incontinence ($n = 9$)^{24,26,30,41,44,50,53,74,117} and parkinsonism ($n = 9$).^{27,29-31,46,57,63,82,119}

TABLE 2 Summary of findings for the most commonly identified prescribing cascades

Initial medication	Suspected ADR	Second medication	Main findings
DH-CCB	Oedema	Loop diuretic	<1 year: aSR 1.46 (95% CI 1.45–1.48); n = 139375 ⁸⁷ <360 days: aSR 1.87 (95%CI 1.84–1.90); n = 55818 ⁷ <360 days: aSR 2.27 (95% CI 1.44–3.58); n = 90 ¹⁰⁵ <360 days: aSR 1.72 (95%CI 1.66–1.78) relative to levothyroxine negative control; aSR 1.45 (1.41–1.49) relative to ACEI/ARB negative control ³⁵ Rate of being dispensed a loop diuretic versus general comparator group ⁵ 1–30 days: aHR 2.51 (95%CI 2.13–2.96) 31–60 days: aHR 2.99 (95%CI 2.43–3.69) 61–90 days: aHR 3.89 (95%CI 3.11–4.87) 91–180 days: aHR 3.20 (95%CI 2.72–3.76) 181–365 days: aHR 2.22 (95%CI 1.90–2.60)
Amiodarone	Hypothyroidism	Thyroxine	<1 year: aSR 3.77 (95% 3.43–4.14); n = 2667 ⁸⁷ <360 days: aSR 3.57 (95%CI 3.17–4.02) ⁵⁴ <1 year: aSR 2.14 (99%CI 1.92–2.39); n = 2613 ⁷⁹ <1 year Australia: aSR 5.30 (95%CI 4.69–5.96); n = 1979 ⁶¹ <1 year Hong Kong: aSR 2.33 (95%CI 1.99–2.72); n = 754 <1 year Japan: aSR 1.77 (95%CI 0.61–5.08); n = 6 <1 year Korea: aSR 1.52 (95%CI 1.29–1.80); n = 657 <1 year Taiwan: aSR 3.26 (95%CI 2.26–4.70); n = 153 <1 year: Pooled aSR 2.63 (95%CI 1.47–4.72) <6 months: aSR 13.6 (95%CI 7.73–25.96) ⁸⁵ <12 months: aSR 12.8 (95%CI 8.44–20.28) <18 months: aSR 11.4 (95%CI 7.98–16.80) <24 months: aSR 11.7 (95%CI 8.32–16.94) <30 months: aSR 10.8 (95%CI 7.86–15.29) <36 months: aSR 10.8 (95%CI 7.89–15.00)
Inhaled corticosteroids	Oral candidiasis	Topical antifungals	<90 days OR 1.66; n = 21 ⁵⁶ <1 year: SR 2.89 (95%CI 2.80–2.97) ³⁹ <1 year: SR 1.94 (95%CI 1.71–2.21) ⁷¹ <360 days: aSR 2.34 (95% CI 2.19–2.50) ⁵⁴
Neuroleptics/ Antipsychotic	Parkinsonian symptoms/ extrapyramidal symptoms	Anti-parkinson medication or Parkinson diagnosis	<90 days: aOR 5.4 (95%CI 4.8–6.1) ¹⁹ <1 year (1 antipsychotic): aSR 9.24 (7.35–11.8); n = 817 ¹⁰⁰ <1 year (2 antipsychotics): aSR 22.2 (9.94–61.7); n = 137 <1 year (\geq 3 antipsychotics): aSR 34.8 (5.87–1413.8); n = 37 Never use: aOR 1.0 (referent); n = 10714 ¹¹⁹ Very-late use (\geq 181 days): aOR 1.1 (95%CI 0.6–1.8); n = 61 Late use (31–180 days): aOR 2.0 (95%CI 1.2–3.3); n = 94 Early use (8–30 days): aOR 6.0 (95%CI 2.3–15.9); n = 43 Current use (\leq 7 days): aOR 3.0 (95%CI 1.7–5.4); n = 80 Typical: aOR 6.4 (95%CI 1.4–28.2); n = 17 Haloperidol: aOR 4.3 (95%CI 0.9–20.1); n = 12 Atypical: aOR 2.4 (95%CI 1.2–4.9); n = 56 Quetiapine: aOR 0.9 (95%CI 0.4–2.2); n = 26 Risperidone: aOR 13.5 (95%CI 1.8–102.1); n = 23 Combined use: aOR 3.2 (95%CI 0.6–17.9); n = 7 Typical antipsychotics: aHR 1.30 (95%CI 1.04–1.58) versus atypical antipsychotic use ²⁹ No therapy: aHR 0.40 (95%CI 0.29–0.43)
Acetylcholinesterase inhibitors	Urinary incontinence	Drugs for urinary frequency and incontinence	During follow-up (1st June 1999–31st March 2003): older adults dispensed acetylcholinesterase inhibitors had a higher risk of subsequently receiving an anticholinergic medication to treat urge urinary incontinence (aHR, 1.55, 95% CI, 1.39–1.72) ²⁶ Donepezil → Medication for managing Lower Urinary Tract Symptoms (LUTS) ⁵³ <3 months: 1.32 (95%CI 1.00–3.50); n = 243 <12 months: aSR: 1.98 (95%CI 1.57–2.50); n = 319 <6 months: aHR 1.47 (95%CI 1.17–1.86) versus memantine users ⁴⁴ <12 months: aHR 1.41 (95%CI 1.17–1.69) versus memantine users Donepezil: aHR 1.55 (95%CI 1.31–1.83) versus rivastigmine use ²⁴ Galantamine: aHR 1.17 (95%CI 0.87–1.58) versus rivastigmine use

(Continues)

TABLE 2 (Continued)

Initial medication	Suspected ADR	Second medication	Main findings
Metoclopramide	Parkinsonian symptoms	Levodopa	<90 days aOR 3.04 (95%CI 2.22–4.17) ²⁰ <90 days aOR 2.94 (95%CI 2.35–3.67) ³¹ Anti-Parkinson medication or diagnosis <1 year: aOR 2.7 (95%CI 1.8–4.1); n = 121 ¹¹⁹
ACE inhibitors	Cough	Antitussive	<1 year OR = 1.58 (95%CI 1.21–2.07) ⁷³ <6 months: SR 2.0 (95%CI 1.8–2.2); n = 1898; estimated 13.4% mistreated cough ¹⁸ <1 year: aSR 1.33 (95% CI 1.31–1.34); n = 141924 ⁸⁷
NSAID	GI symptoms	Anti-ulcer medication	<4 weeks: aSR 2.50 (95%CI 2.27–2.76); n = 2016 ¹³² <100 days: aSR 1.80 (95%CI 1.64–1.99); n = 1814 ⁸⁹ <1 year: SR 1.71 (95%CI 1.67–1.74); n = 49646 ¹⁰¹
Ranitidine	Heart failure	Furosemide	<1 year: aSR 1.08 (95%CI 1.04–1.12); n = 10875 ⁷⁵ <1 year: aSR 1.24 (95% CI 1.17–1.31); n = 5554 ¹⁰⁶
Rosiglitazone	failure	Furosemide	<1 year Australia-1: aSR 1.70 (95%CI 1.34–2.15) ⁹⁶ <1 year Australia-2: aSR 1.63 (95%CI 1.51–1.76) <1 year Canada: aSR 1.65 (95%CI 1.57–1.73) <1 year Pooled estimate (Australia & Canada): aSR 1.65 (95%CI 1.58–1.72) <1 year Hong Kong: aSR 3.37 (95%CI 1.69–6.72) <1 year Korea: aSR 1.14 (95%CI 1.08–1.21) <1 year Taiwan: aSR 1.12 (95%CI 0.99–1.25) <1 year Pooled estimate (Asia): aSR 1.21 (95%CI 1.01–1.45) July 2000–December 2007: aSR 1.73 (99%CI 1.34–2.24) ¹¹³
SGLT2-I	Genital infections	Antifungal	<30 days: aSR 1.35 (95%CI 1.26–1.44) ⁴⁷ <60 days: aSR 1.48 (95%CI 1.40–1.56) <90 days: aSR 1.53 (95% CI 1.43–1.60) <180 days: aSR 1.42 (95%CI 1.37–1.47) <365 days: aSR 1.24 (95%CI 1.20–1.28) Genital infection occurred more frequently among SGLT2-I users than DPP-4 users (2.9% vs, 0.9%, aHR 3.50, 95%CI 1.95–5.89) ⁸⁸
DOAC	Depression	Antidepressant	<3 months: aSR 1.29 (95%CI 1.23–1.35); n = 7253 ⁹³ <6 months: aSR 1.28 (95%CI 1.24–1.33); n = 12530 <12 months: aSR 1.26 (95%CI 1.23–1.30); n = 20613 SSRI <6 month: SR 1.57 (1.37–1.77); n = 1137; NNTH 202 ¹⁰² Other antidepressant <6 month: SR 1.59; 1076; (1.41–1.80); NNTH 207
High ceiling diuretics	Hypokalaemia	Potassium	Furosemide <360 days: aSR 2.94 (95% CI 2.83–3.05) ⁵⁴ High ceiling diuretic <1 year: SR 3.31 (95%CI 3.24–3.38); n = 48539 ¹⁰¹
Statins	Lower urinary tract symptoms (LUTS)	Drugs for urinary frequency and incontinence	<91 days: aSR 1.21 (95% CI 1.00, 1.46); n = 446 ⁵⁰ <182 days: aSR 1.19 (95% CI 1.04, 1.38); n = 785 <365 days: aSR 1.17 (95% CI 1.05, 1.30); n = 1373 <1 year: aSR 1.17 (95% CI 1.16–1.19); n = 107422 ⁸⁷
Statins	Skin soft tissue infection	Antibiotic (Dicloxacillin or Flucloxacillin)	<1 year: aSR 1.18 (95% CI 1.15–1.22); n = 23068 ⁸⁷ <91 days: aSR 1.40 (95%CI 1.29–1.52); n = 2498 ⁷⁶ <182 days: aSR 1.41 (95%CI 1.33–1.50); n = 4277 <365 days: aSR 1.40 (95%CI 1.34–1.47); n = 7726
Statins	Depression	Antidepressant	<1 year: aSR 1.19 (95% CI 1.18–1.21); n = 117443 ⁸⁷ Simvastatin → Antidepressant (April 1991–December 1995): aSR 1.59 (1.08–2.45); n = 91 ⁹⁸
Statins	Muscle cramps	Quinine	<360 days: aSR 1.69 (95% CI 1.61–1.77) ⁷⁰ <1 year: aSR = 1.16 (95%CI 1.04–1.29); n = 1326 ⁵¹
Brinzolamide	Heart failure	Furosemide	<1 year Brinzolamide: aSR 1.18 (95%CI 1.06–1.32); n = 1304 ⁷⁵ <1 year Brinzolamide: aSR 1.37 (95% CI 1.16–1.62); n = 564 ¹⁰⁶
Latanoprost	Heart failure	Furosemide	<1 year Latanoprost: aSR 1.11 (95%CI 1.04–1.19); n = 3619 ⁷⁵ <1 year Latanoprost: aSR 1.48 (95% CI 1.38–1.59); n = 3107 ¹⁰⁶
Carbamazepine	Hypothyroidism	Levothyroxine	1998–2004: aOR 1.37 (95%CI 1.13–1.65) ¹³⁰ <1 year: aSR 1.21 (99%CI 1.08–1.34) ⁷⁹

TABLE 2 (Continued)

Initial medication	Suspected ADR	Second medication	Main findings
Valproate	Hypothyroidism	Levothyroxine	1998–2004: aOR 1.72 (95%CI 1.40–2.11) ¹³⁰ <1 year: aSR 1.34 (99%CI 1.20–1.49) ⁷⁹
Lithium	Drug induced tremor Parkinson	Anti-parkinson drug	Jan 1995–December 1999: RR 1.88 (95%CI 1.60–2.20) ⁸² Up to 2 year follow-up (referent valproic acid): aHR 1.50 (95%CI 0.68–3.36) ²⁷ Up to 2 year follow-up (referent antidepressant): aHR 1.56 (95%CI 0.98–2.48)
Lithium	Hypothyroidism	Thyroxine	1998–2004: aOR 1.41 (95%CI 1.14–1.74) ¹³⁰ <360 days: aSR 3.43 (95%CI 2.55–4.70) ⁵⁴
Benzodiazepine	Dementia	Anti-dementia drug	<3 months: aSR 1.24 (95%CI 1.05–1.45); n = 625 ⁶⁶ <6 months: aSR 1.20 (95%CI 1.06–1.37); n = 973 <12 months: aSR 1.23 (95%CI 1.11–1.37); n = 1450 <24 months: aSR 1.34 (95%CI 1.23–1.47); n = 2049 <36 months: aSR 1.41 (95%CI 1.29–1.53); n = 2408 <48 months: aSR 1.44 (95%CI 1.33–1.56); n = 2653 <3 years: aSR 2.19 (95%CI 1.92–2.49); n = 1285 ⁹⁴ <2 years: aSR 2.00 (95%CI 1.71–2.34); n = 780 <1 year: aSR 1.77 (95%CI 1.39–2.27); n = 286
SSRI	Urinary incontinence	Drugs for urinary frequency and incontinence (or incontinence products) ⁴¹	Paroxetine <1 year: aSR 1.77 (95%CI 1.33–2.36) ⁵³ During SSRI (before SSRI as referent): IDR 1.57 (95%CI 1.38–1.79) ⁴¹ During SSRI (after SSRI as referent): IDR 2.03 (95%CI 1.76–2.34) During SSRI (before and after SSRI as referent): IDR 1.75 (95%CI 1.56–1.97) Patients had a 61% higher risk for incontinence (aRR 1.61, 95%CI 1.42–1.82)

Abbreviations: aHR, adjusted hazard ratio; aOR, adjusted odds ratio; aSR, adjusted sequence ratio; IDR, incidence density ratio; NNTH, number needed to har; SR, crude sequence ratio.

3.6 | New medication(s) prescribed

The medication sub-classifications most frequently initiated as a new medication in the 94 studies are summarised in Figure 2A. Seventy-eight studies reported at least one significant positive association, indicating a potential prescribing cascade (Table 1 and Figure 2A–C).

The most commonly identified prescribing cascades are summarised in Table 2. These include; amiodarone associated with subsequent thyroid hormone prescriptions for hypothyroidism ($n = 5$),^{54,61,79,85,87} CCBs associated with diuretic prescriptions to treat peripheral oedema ($n = 5$),^{5,7,87,104,105} topical antifungals to treat oral candidiasis following inhaled corticosteroids ($n = 4$),^{39,54,56,71} anti-Parkinson medication to treat Parkinsonian symptoms following antipsychotic initiation ($n = 4$),^{19,29,100,119} urinary anticholinergics to treat urinary incontinence following acetylcholinesterase inhibitors ($n = 4$),^{24,26,44,53} and antitussives to treat cough following angiotensin-converting enzyme inhibitors (ACEIs) ($n = 3$).^{18,73,87} Additional prescribing cascades identified included metoclopramide to anti-Parkinson medication ($n = 3$),^{20,31,119} and NSAID to anti-ulcer medication.^{89,91,101}

No association between drug pairs could be determined for several studies, largely due to either a cross-sectional study design examining concurrent drug use, insufficient drug-pair sample size to determine a sequence ratio or reporting of

incidence rates with no incidence rate ratio (labelled N/A in Figure 2).^{6,21,25,30,43,46,53,55,63,64,78,79,112,114,115} Several studies reported at least one negative association between drug pairs, indicating a reduced likelihood of the second medication being initiated (see eTable 3 Appendix S1).^{33,60,68–70,74,81,87,89,93,111}

3.7 | Modifiers of identified associations

Older people (aged ≥ 65 years) were more likely to receive; (i) anticholinergics for urinary incontinence following SSRI initiation,⁴¹ (ii) ulcer drug therapy within 100 days of NSAID initiation,⁸⁹ (iii) diuretic to treat beta-blocker induced oedema,³⁶ and, (iv) thyroxine for hypothyroidism following amiodarone initiation.⁸⁵ Females were more likely to receive an antitussive for cough following ACEI initiation,⁷³ anticholinergic medication for urinary incontinence following acetylcholinesterase inhibitor^{24,30} and SSRI initiation,⁴¹ and levothyroxine following amiodarone initiation.⁸⁵

Differential associations were identified for initial medication dosage in nine studies. Those who received higher doses of CCBs^{5,7} and gabapentinoids were more likely to receive a diuretic for oedema;²⁸ higher doses of inhaled corticosteroids were associated with a greater likelihood of treatment for oral candidiasis;³⁹ and higher metoclopramide dosage was found to increase the likelihood for dopaminergic treatment initiation.²⁰

Polypharmacy (≥ 5 drugs) was associated with a greater likelihood of receiving thyroid hormones for amiodarone induced hypothyroidism.⁸⁵

3.8 | Intentional and unintentional cascades

The intentionality of potential prescribing cascades was not reported in any study nor was the intended duration (if any) of the prescription of the second medication. One study provided a breakdown of prescriptions for the initial drug by prescriber type: 23% private cardiologist, 35.5% hospital practitioner, 30.3% General Practitioner, and 11.3% other private specialist, but did not provide details of the prescriber of the second drug.⁹³ Another study reported that of the sample who initiated the second drug (irrespective of initiating the first drug), 87.1% of prescriptions were started by family physicians.⁵¹

FIGURE 3 Quality appraisal summary of included studies ($n = 98$): (A) cohort studies; (B) case-control studies; (C) cross-sectional studies.



3.9 | Clinical importance of prescribing cascade

Two studies reported a number needed to harm (NNTH) for investigated cascades.^{62,102} (See Table 1). One study ($n = 90$) conducted a medical chart validation study of those initiated a loop diuretic after initiating a dihydropyridine CCB ($n = 64$) and determined that 54.7% ($n = 35$) experienced a prescribing cascade.¹⁰⁵

3.10 | Quality assessment

Overall, the methodological quality varied across included studies (Figure 3 and eTables 4–6, Appendix S1). Among the retrospective cohort studies (eTable 4, Appendix S1) there was a lack of clarity surrounding the similarity of exposed and unexposed groups at baseline and the presence of the outcome at the start of the study. For case-control studies (eTable 5, Appendix S1), reporting of baseline

comparison of cases and controls was inadequate as well as the appropriateness of matching cases with controls.

4 | CONCLUSION AND IMPLICATIONS

4.1 | Principal findings

This systematic review identified 101 studies across 103 publications that examined potential prescribing cascades across a broad range of pharmacological drug groups. All studies used routine administrative data that included either medication prescribing or dispensed medications data. Of the 101 included studies, 78 (77%) reported at least one significant positive quantitative association that indicates a potential prescribing cascade. The most commonly identified prescribing cascades include: (i) CCBs → loop diuretics to treat peripheral oedema ($n = 5$); (ii) amiodarone → thyroxine to treat hypothyroidism ($n = 5$); inhaled corticosteroids → topical antifungal to treat candidiasis ($n = 4$); (iii) antipsychotics → anti-Parkinson medication to treat Parkinsonism ($n = 4$); and (iv) acetylcholinesterase inhibitors → drugs for urinary frequency ($n = 4$).

Study methodological quality was variable with a considerable proportion of studies not reporting participant demographics. Almost two-thirds of included studies used PSSA methodology in which all included participants have experienced the outcome at the start of the study. A recent scoping review reported that whilst the PSSA method is a useful tool in detecting prescribing cascades, such cascades need careful clinical review as there is a risk of both false positive and false negative findings.¹²⁰ This is particularly problematic when screening for cascades without predefined hypotheses. In our systematic review, the vast majority of included studies ($n = 94$, 93%) examined predefined medications as potentially contributing to a prescribing cascade. However, PSSA analyses cannot determine causality and should be interpreted with caution.

Several well-designed cohort and case-control studies examining prescribing cascades were identified. For example, a Canadian population-based study reported that incident CCB users had a higher cumulative incidence of loop diuretic use at one year follow up compared to patients dispensed ACEIs or angiotensin-II-receptor blocker antihypertensives (adjusted hazards ratio 1.4% vs. 0.7%, $p < 0.001$).⁵ In a US case-control study, metoclopramide users were three times more likely to begin use of a levodopa-containing medication compared with nonusers (OR = 3.09; 95% CI 2.25 to 4.26).²⁰ Risk increased with increasing daily metoclopramide dose and the effect persisted after adjustment for demographic, health service utilization, and medication use variables.²⁰

Fifteen of 101 studies focused specifically on older populations, with 11 reporting a significant association between increasing age and prescribing cascade occurrence. Older adults are more likely to experience medication-related harm due to increasing prevalence of multimorbidity, polypharmacy and age-related physiological changes that affect drug metabolism.^{9,10,121-123} Furthermore, ADRs are more difficult to diagnose in older adults due to their often non-specific

presentation and overlap with pre-existing conditions or conditions likely to develop among older adults.^{1,8,124}

4.2 | Comparison with existing literature

Two scoping reviews of prescribing cascades have been conducted to date, one that focused on literature surrounding the prevention, detection and reversal of prescribing cascades¹¹ and the second that focused on the use of PSSA as a potential pharmacovigilance tool.¹²⁰ In 2018, Brath et al retrieved 10 original investigations and seven case reports pertaining to prescribing cascades.¹¹ A considerable number of studies have been published since, indicating that this is a rapidly developing field. Morris et al. concluded that PSSA methodology demonstrated only moderate sensitivity and specificity in identifying prescribing cascades and more consistency was required in how these studies were reported.¹²⁰ As described previously, similar issues with methodological quality were identified in this systematic review.

4.3 | Clinical and research implications

Multi-country studies have shown variation in prescribing cascade likelihood both within and across countries,^{60,95,96} underscoring the need to consider the local prescribing context. Differences in sample demographics, medication availability, approved clinical indications, help-seeking behaviour and prescribing cultures or genetic polymorphisms may influence the incidence of prescribing cascades.

The complexity of optimising prescribing for patients with multimorbidity presents challenges for the prescriber due to the preponderance of single-disease guidelines, resultant polypharmacy, fragmentation and lack of continuity of care and resourcing constraints.¹²⁵ Identification of ADRs remains a clinically challenging area, particularly in relation to older adults. Non-specific presentation of ADR symptoms in older adults, such as delirium, falls, fatigue and constipation, can be challenging to identify as being medication-related as such symptoms have several causes and may overlap with existing multimorbidity.^{8,124} The failure to recognise an ADR may result in a prescribing cascade, furthering the risk for additional medication-related harm.^{1,2} The potential for ADRs should be considered as part of the differential diagnosis for all patients reporting new symptoms, particularly among those who have started a new medication within the previous year.^{1,8,124}

Developing an explicit list of evidence-based prescribing cascades is one way of supporting clinicians' awareness and detection of this issue. The iKASCADE international consortium are currently developing an inventory of prescribing cascades affecting older adults, through a modified Delphi procedure where international experts in medicines management for older adults will rank a list of prescribing cascades as to their clinical importance.¹²⁶ The development of an explicit list of clinically important and common prescribing cascades is an important step in raising awareness of this

issue and in supporting clinicians to detect cascades.¹²⁷ To maximise use in clinical practice will require explicit criteria of prescribing cascades be incorporated into existing electronic health record and prescribing support systems. Such systems will need to be able to detect the sequential prescription of drugs known to represent potentially inappropriate prescribing cascades.¹²⁷

The use of routine administrative data in included studies means that information on the broader clinical context and the rationale for medication prescribing is lacking. The identification of significant negative associations between drug pairs may indicate that prescribers are aware of certain prescribing cascades and proactively avoid their development or that therapeutic alternatives were prescribed. However, no exploration of intentionality of identified cascades could be made based on the data used in included studies.

Overall, it is difficult to determine the clinical importance of prescribing cascades identified as few studies examined clinical endpoints.^{48,62,102} One study examined the association between prescribing cascades that resulted in prochlorperazine initiation and reported a subsequent 49% increased risk of hip fracture.⁴⁸ Future research is required to determine the relative clinical impact of increased medication exposure and the clinical appropriateness of prescribing cascades.

4.4 | Strengths and limitations

This systematic review extends the work of previously published scoping reviews^{11,120} by conducting a comprehensive literature search using several databases, including several grey literature searches.

This study also has some limitations. The lack of a MeSH term for prescribing cascades meant broad search terms were used, which led to a high yield of citations to be searched. Additional information was sought from study authors but a small number of studies ($n = 10$) could not be retrieved for eligibility assessment due to the lack of access to the full text or a translated version. The information collated is somewhat limited by the methodological and reporting quality of included studies.

5 | CONCLUSION

Prescribing cascades are of increasing interest to the research and clinical communities, with a broad range of medications involved. The identification of the most common prescribing cascades can support optimising prescribing as one part of identifying potentially inappropriate prescribing. Few studies have examined the clinical importance or the broader clinical context, including intentionality of prescribing cascades, thereby limiting the inferences that can be drawn about the implications for clinical practice. Challenges remain in differentiating ADR symptoms from that of new onset disease and advancing age and frailty.^{1,8,124} ADRs should be considered as part of the differential diagnosis in patients presenting with new

symptoms, particularly for those who have started a new medication in the preceding 12 months.

AUTHOR CONTRIBUTIONS

Conception and funding acquisition: EW. Study design EW, AD, FM, FB, BC, SK, and TF. Data acquisition: AD, FS, and EW. Data interpretation: AD, FS, TD, FM, FB, BC, TF, SK, and EW. Drafting of manuscript: AD, and EW. Revising of manuscript and agreement of final manuscript: AD, FS, FM, FB, BC, SK, TF, TD, and EW.

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CONFLICT OF INTEREST

The authors have no conflicts of interest to declare.

DATA AVAILABILITY STATEMENT

Additional systematic review data is available from the authors on request.

ETHICS STATEMENT

Ethical approval was not required for this systematic review.

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SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

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