Real World Experience of Alcohol and Tobacco in a National Health System: The Veterans Aging Cohort Study

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COI

• I have funding from multiple federal sources including several NIH institutes. (mainly NIAAA) and the Veterans Health Administration

The opinions expressed are my own

Sample size required

Self reported measures of exposure

Time delay between exposure and cancer presentation

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Sample Size: U.S. Veterans Healthcare System an Exceptional Real-World Laboratory

- Largest integrated healthcare system in USA
 - 827 community-based outpatient clinics
 - 151 hospitals/medical centers
 - 126 nursing homes
 - 13.2 million in care, ~ 6 million annually
- National, paperless, EMR since 1998
- Largely captured, longitudinal aging population
- Cancer registry
- Smoking and alcohol use data
- Pharmacy fill data
- Excellent mortality ascertainment
- Links to
 - National Death Index
 - Medicare and Medicaid data
 - Million Veteran Study genetics/omics for >1 million



Sample Size: VACS EHR-based Cohorts

- Full national sample
- N~13.2 million
- All Veterans receiving care in the VA system
- HIV study
- 60,000 PWH, 120,000 PWoH
- 1 PWH:2 PWoH demographically-matched

VACS-National



VACS-HIV



- 4,352 PWH; 4,561 PWoH
- Consented survey substudy
- 1 PWH:1 PWoH
- 9 VA sites
- Patient & provider surveys
- Linkage to EHR

VACS-HIV Survey



- 1,730 PWH, 926 PWoH
- Consented biomarker substudy
- 2 PWH:1 PWoH
- Banked specimens, select labs, DNA
- Linkage to EHR + Survey

VACS-HIV Biomarker



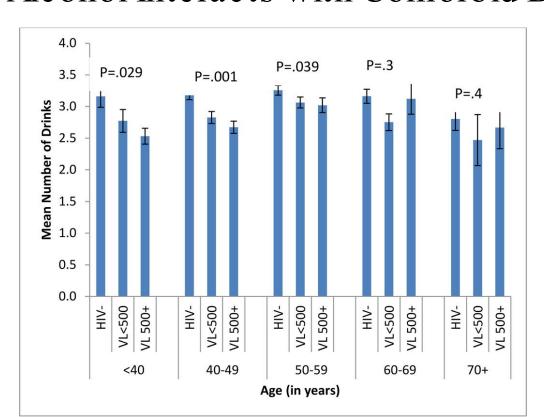
- 534 PWH at VACS Survey Sites
- Consented biomarker substudy
- Tripartite with KPNC (284 PWH), SHCS (451 PWH)
- Banked specimens, select labs, DNA
- Linkage to EHR + MASH specific patient survey and pharmacist

MASH





Alcohol Interacts with Comorbid Diseases Like HIV



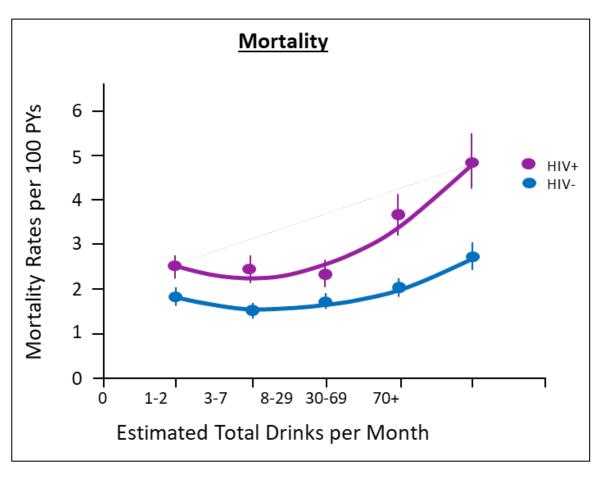
Drinks to Feel a "Buzz"





David Feillen

en Kathy McGinnis



McGinnis, et al. AIDS Behav (2016). PMID: 26936030 Justice et al. DAD (2016). PMID: 26861883

Sample size required

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Time delay between exposure and cancer presentation

Self Reported Smoking

Self report <u>may be</u> subject to under reporting

- Use decreases with age (Figure)
- Social desirability bias
- Abstainer bias

Cotinine gold standard (Table)

- Only measures recent use
- ICD codes terrible
- Self report (survey or EHR) good

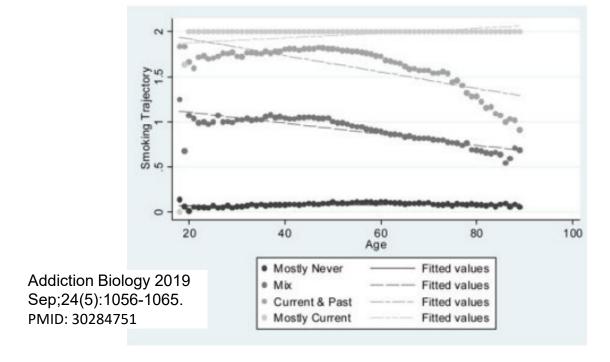


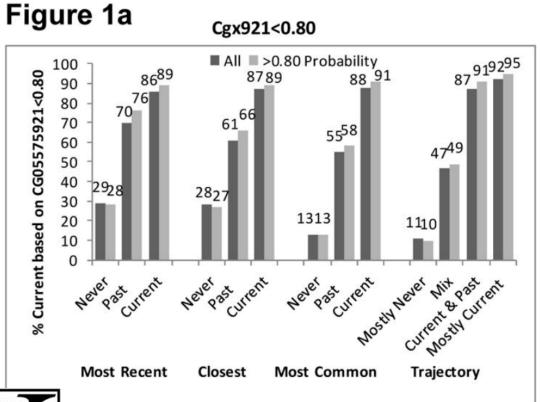
Table 2. Smoking from clinical reminder, survey, and ICD codes compared with cotinine (≥30 and ≥10) as reference standard (n=318), FY 2019

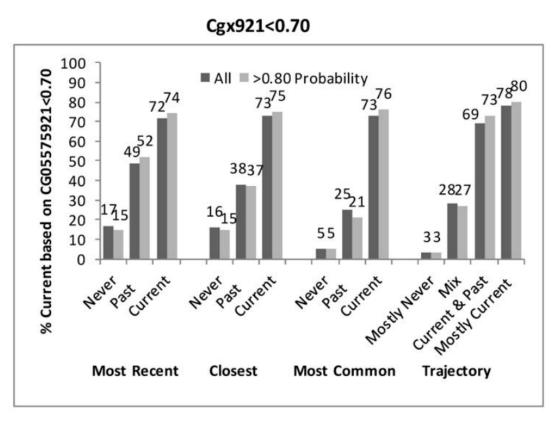
	Percent						
	Prevalence	Agreement	Sensitivity	Specificity	PPV	NPV	Kappa
Cotinine ≥30	42		_	_	_	_	_
Clinical reminder	39	91	86	95	92	90	.81
Survey	38	92	85	97	95	90	.83
ICD-10	24	77	51	97	92	73	.50
Cotinine ≥10	46		_	_	_	_	_
Clinical reminder	39	90	82	97	95	86	.79
Survey	38	91	82	99	98	86	.81
ICD-10	24	74	48	97	93	69	.46

ICD: International Classification of Disease.



Methylation Sites May Be an Option for Unbiased Measurement





Measuring Alcohol Use--Even Trickier

Self report subject to under reporting

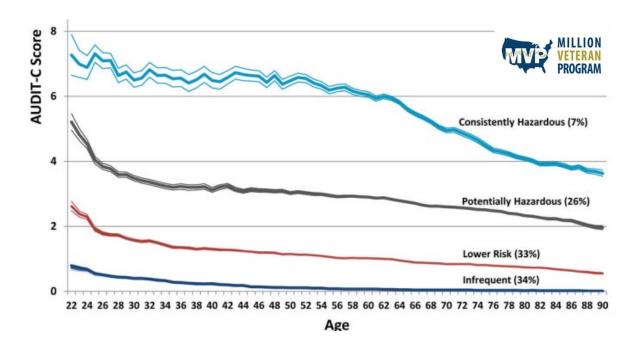
- Social desirability bias
- Use decreases with age (Figure)¹
- Abstainer bias worse with age

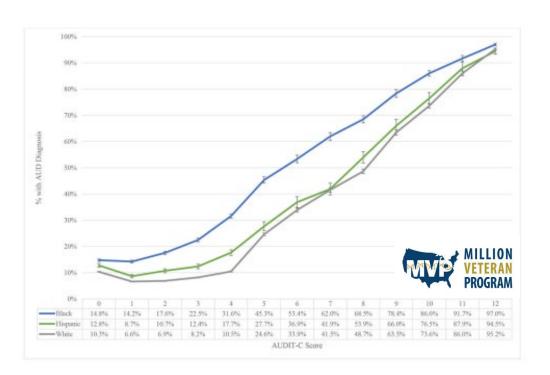
Phosphatidyl ethanol (PEth) gold standard

- Only recent use (past 21 days)
- Demonstrates differential social desirability bias²
 - Among PWH, highest mortality among those reporting abstinence and PEth ≥8 ng/mL (5.69 per 100 PY, 95% CI 3.78–8.56)
 - Comparable with high-risk AUDIT-C and PEth ≥8 ng/mL (6.12 per 100 PY, 95% CI 3.82– 9.85)
 - Results similar after full adjustment and among PWoH

ICD Codes demonstrate racial bias³

- 1. Alcohol Clin Exp Res. 2019 March; 43(3): 465-472.
- 2. J Acquir Immune Defic Syndr. 2018 February 01; 77(2): 135-143.
- 3. Am J Psychiatry. 2023 June 01; 180(6): 426-436.





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Time Delay Between Exposure and Cancer: Risk Takes 10-20 Years to Resolve



HCC Risk Among Never Drinkers Vs. Just Quit Drinking

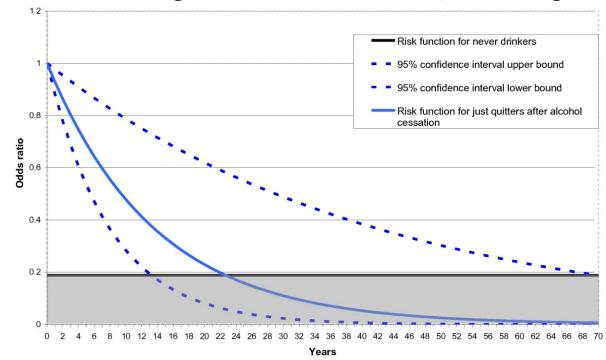


Illustration: Risk decline of liver cancer post cessation of alcohol consumption compared to just quitters.

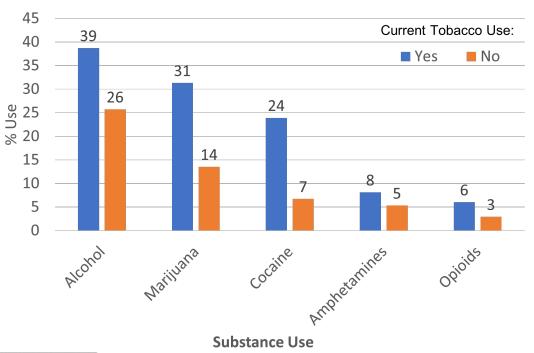
BMC Cancer 11, 446 (2011)

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Use of One Substance Increases Use of Other Substances



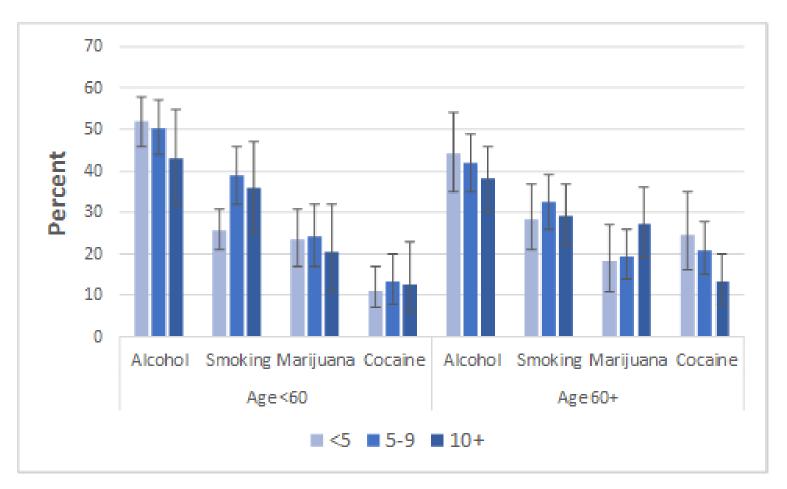
Association of tobacco use with other substance use

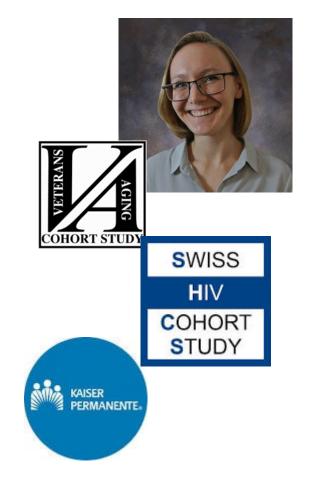
		Odds Ratio (95% CI)		
	N (%)	Unadjusted	Adjusted*	
Peth (>20 ng/mL)	184 (38.7%)	1.86 (1.46, 2.37)	1.80 (1.40, 2.33)	
Cannabis	147 (31.3%)	2.93 (2.20, 3.89)	3.62 (2.65, 4.95)	
Cocaine	112 (23.9%)	4.34 (3.05, 6.17)	3.84 (2.63, 5.60)	
Opioids	28 (6.0%)	2.16 (1.22, 3.83)	2.42 (1.32, 4.44)	
Amphetamines	38 (8.1%)	1.57 (0.99, 2.48)	1.57 (0.96, 2.57)	

^{*}Models adjusted for age, gender, race and ethnicity, and cohort



Substance Use and Polypharmacy Co-Occur Among PWH





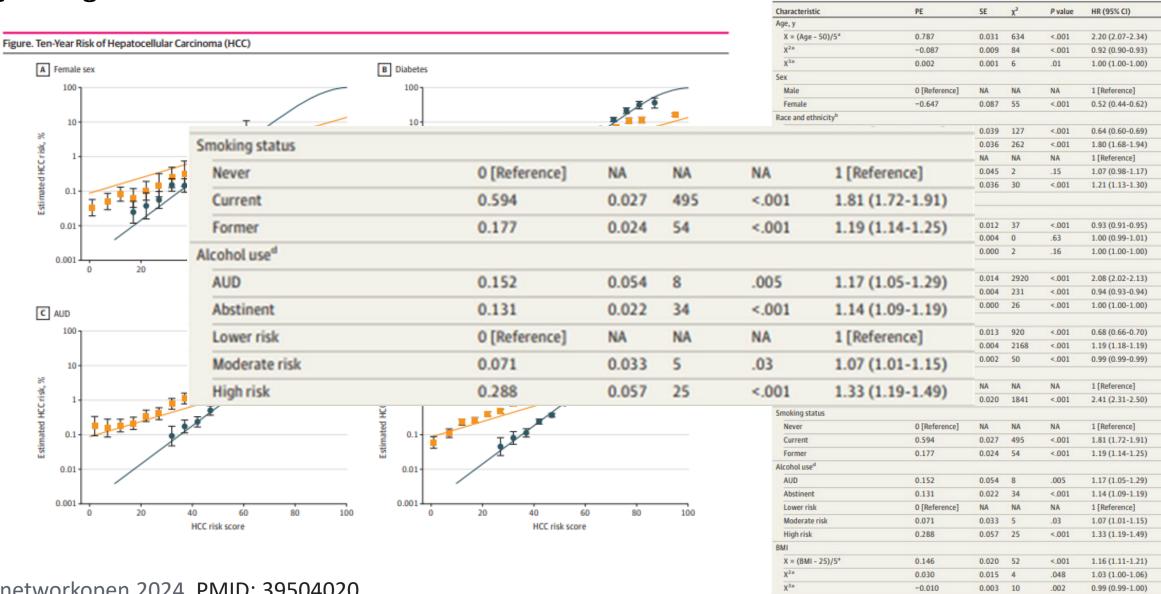
- Medications, Alcohol, Substance use in HIV Study (MASH)
- Extends VACS to include Kaiser Permanente Northern California and Swiss Cohort
- Prospectively collect data on: all medications, Self report and biomarker verified substance use

Alcohol and Smoking Independently Associated with Hepatocellular Carcinoma

Table 3. Cox Proportional Hazards Regression Model Fit to Development Sample of Veterans

With Hepatocellular Carcinoma Events During a Maximum 10-Year Follow-Up

Adjusting for Liver Disease in >6.5 M Veterans



jamanetworkopen.2024. PMID: 39504020.

Glucagon-like peptide-1 receptor agonists but not dipeptidyl peptidase-4 inhibitors reduce alcohol intake

Mehdi Farokhnia,¹ John Tazare,² Claire L. Pince,¹ Nicolaus Bruns Vi,¹ Joshua C. Gray,³ Vincent Lo Re III,⁴ David A. Fiellin,⁵ Henry R. Kranzler,⁶ George F. Koob,⁷ Amy C. Justice,⁵ Leandro F. Vendruscolo,⁸ Christopher T. Rentsch,⁹ and Lorenzo Leggio¹

Mouse and rat models for unhealthy alcohol intake: Previously showed effect of the GLP-1RA (semaglutide) Tested two DPP-4Is (linagliptin and omarigliptin)

Compared changes in AUDIT-C between propensity-score-matched recipients of GLP-1RA, DPP-4I, & unexposed

GLP-1RA recipients had a greater reduction in AUDIT-C scores than unexposed and DPP-4I recipients

No differences between DPP-4I and unexposed

n=27,231 GLP-1RA vs. unexposed 77,911 DPP-4I vs. unexposed 28,996 for GLP-1RA vs. DPP-4I

Journal of Clinical Investigation, Published Online March 6, 2025 PMID: 40048376

EMR Cohorts Address Challenges in Understanding the Role of Alcohol and Tobacco in Cancer

Sample size

- 9 m Veterans/yr
- Other health systems

Self report

- Serial PEth
- Methylation Data

Time delay

- 20 yrs of data
- Need biomarkers/ intermediate outcomes

- Propensity adjustment
- Causal modeling

Consider a More Personalized Clinical Approach



- Get a PEth first, then, if needed address alcohol
- Use EHR decision support to create personalized messages accounting for other factors (cancer, heart disease, drug interactions)
- Substitute a positive behavior (exercise) for use of substances—especially when they are used to self medicate
- Select medications that address multiple modifiable factors (GLP-1 inhibitors for 3) but consider whether life long treatment is reasonable





VACS

Veterans Aging Cohort Study

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