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1 **Regional Economic Potential for Recycling Consumer Waste Electronics in the United**
2 **States**

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7
8 **Abstract**

9 **Waste electronics are a growing environmental concern, but also contain materials with**
10 **great economic value. If properly recycled, waste electronics could enhance the**
11 **sustainability of vital metal supply chains by offsetting the increasing demand for virgin**
12 **mining. However, rapid changes in size and composition of electronics complicate their end-**
13 **of-life (EOL) management. Herein, we couple material flow and geospatial analyses on over**
14 **90 critical consumer electronic products and find that over 1 billion devices, representing**
15 **up to 1.5 million tonnes of mass, could be discarded annually in the United States (U.S.) by**
16 **2033. Emerging electronics such as connected home, health, and AR/VR devices have**
17 **become the fastest-growing types in the waste stream. We highlight policy opportunities to**
18 **develop various sustainable circularity strategies around metal supply chains, by showing**
19 **the potential to integrate electronic waste and virgin mining pathways in Western U.S.**
20 **regions, while new infrastructure designed specifically for waste electronics treatment is**
21 **favorable in the Central and Eastern U.S. Further, we show the importance of building**
22 **national-level refining and tear-down databases to improve electronics EOL management in**
23 **the next decade.**

24 Due to the growing consumption of electronic equipment, and the relatively short lifetimes of
25 many of these devices, the amount of discarded electronics within the waste streams (waste
26 electronics) is rapidly increasing. Globally, the annual waste electronics generated has grown
27 from less than 25 million tonnes (Mt) per year in 2009 to around 53.6 Mt in 2019.¹⁻³ The number
28 of waste electronics generated in the United States (U.S.) in 2019 was 6.9 Mt (approximately
29 12.9% of world generation), corresponding to a generation per capita of around 21 kg, which is
30 approximately three times the world's average value (7.3 kg).³

31 Only 17% of the waste electronics generated globally is properly recycled and the recycling rate
32 in the U.S. is around 14% to 40% depending on the reporting agency and selected types of
33 electronics.^{3,4} Domestic waste electronics that are exported for recycling are hard to regulate,
34 owing to the different levels of maturities in terms of recycling technology and policy incentives
35 between countries.^{5,6} Due to the high heterogeneity between types and brands, waste electronics
36 are a special stream of municipal solid waste that requires complex considerations during end-of-
37 life (EOL) management.^{7,8} Without safe and effective EOL processes and government oversight,
38 which is the case for several developing nations, the receiving regions face various environmental
39 pollution and public health concerns such as heavy metal poisoning and food chain
40 contamination.^{5,9-11}

41 However, with proper EOL management, waste electronics can have positive socio-economic
42 impacts as they contain materials of significant value that could allow proper recycling to have
43 economic benefits. In terms of overall material composition, waste electronics are a combination
44 of plastics, metals, etc. The current waste electronics recycling process includes preprocessing
45 and metal recycling. A general waste electronics device includes plastic covers, and printed
46 circuit boards (PCBs) that hold the various electronic parts (i.e., resistors, capacitors), batteries,
47 etc.¹² Each of these components offers unique value upon recycling. Existing techno-economic

48 analysis (TEA) studies have demonstrated the potential of making profits via recovering metal
49 resources, particularly gold, from the PCBs being the main source of revenue.¹³⁻¹⁵

50 Therefore, while improvements in international regulations of waste electronics recycling could
51 be beneficial, it is also critical to predict and assess potential waste management strategies within
52 the U.S. around metal recovery from waste electronics to identify favorable economic and
53 environmental pathways, obviate future resource scarcities, and create a more circular economy
54 to increase the resiliency of domestic material supply.

55 Within waste electronics, small to mid-size consumer electronics (i.e., smartphones, fitness
56 devices, connected home devices, AR/VR equipment, drones, and computers) represent an
57 emerging stream in recent years that are heavily affected by the evolvement of technology
58 development and consumer behaviors. Material flow analysis (MFA) is an effective and strategic
59 way of predicting the quantity and composition of waste materials based on information such as
60 historical sales and possession data, which has been applied to model waste electronic generation
61 in different parts of the world.¹⁶⁻¹⁸ However, existing MFAs on waste electronics in the U.S. has
62 focused mostly on the temporal changes of common and relatively conventional small-to-midsize
63 waste electronics (i.e., TVs, mobile phones, computers, and monitors) within the past decade.¹⁹⁻²¹

64 Due to potential challenges such as data availability at the time of these studies and the rapid
65 change in consumer behaviors, limited research has been conducted regarding the impacts of the
66 emerging small-to-midsize electronics that have been introduced in recent years. Also, the
67 variance of the metal compositions for different types of electronics is rarely taken into account.

68 Lastly, existing research on the geospatial modeling of waste electronics recycling reports is
69 limited to selected waste electronics in selected states.²²

70 Based on these knowledge gaps, herein we couple geospatial analysis on top of state-of-the-art
71 MFA to comprehensively capture the temporal variations and predict future trends regarding the
72 amount, composition, and potential value within these consumer waste electronics generated in

73 the U.S. We also predict how the newly-developed, emerging electronic products can potentially
74 reshape the resource availability from waste electronics, and how these changes will potentially
75 affect the future metal refining infrastructure in the U.S.

76 Since previous analyses showed that gold holds most of the metallic values (up to or higher than
77 80%) within the waste electronics,^{13,14,23} in this analysis we use gold as the primary indicator to
78 examine the economic potential (Fig. 1). The MFA results are used in subsequent geospatial
79 models to characterize the spatial distribution of waste electronics and the embedded value within
80 the U.S. This study identifies the underlying potential connections between gold recovery and the
81 current metal refining industries within the U.S. and offers recommendations on creating an
82 electronics-centered circular economy for different future scenarios. Other secondary resources,
83 such as base metals (i.e., copper, aluminum), and plastics can also add additional value to the
84 recycling of electronics waste, which are recognized and discussed as well (Fig. 1).

85

86 **Results**

87 *Temporal and spatial distribution of waste electronics*

88 Fig. 2a shows the modeled number of waste electronics in the next decade, based on the historical
89 sales data and their predicted lifespans. With the fast growth scenario, the amount of small to
90 mid-size waste electronics produced in 2033 could exceed 1 billion units, and the slow growth
91 will lead to approximately 700 million units. In terms of mass, waste electronics generated in the
92 U.S. are estimated to reach from 1.2 to 1.5 million tonnes (Fig. 2b), with the shift toward smaller
93 electronics partially offsetting the growth in the number of units. Note that in this study, heavy
94 consumer appliance electronics (i.e., fridges or ovens) are not included therefore, both the number
95 and mass are conservative estimates.

96 Figs. 2c and 2d show the mean and general growth trend of representative resources in waste
97 electronics over the next decade, within which PCBs contain the most valuable materials
98 recoverable, including precious metals such as gold (Fig. 2 c).^{13,14} The steady trend of gold
99 availability from waste electronics in recent years (i.e., 2017-2021) is consistent with previous
100 studies. For example, Althaf et al. 2021²¹ and Golev et al. 2016²⁴ both estimated a steady and
101 slow decline in gold availability from 2015-2018 among conventional electronics in U.S. and
102 Australia, respectively, owing to a trends towards smaller electronic devices and more efficient
103 use of gold within those electronics.

104 However, existing tear-down studies on emerging smart devices, which often contain more
105 complex electronics (i.e., smartphones, and tablets) have shown higher PCB and gold content
106 than that of larger, stationary, and wired electronics (e.g., DVDs and VCRs)²⁵. Even though the
107 total amount of gold used in electronics has declined in recent years, which influences the
108 economics of recovery, the number of more complex devices with higher PCB and gold content
109 has been increasing. These tradeoffs indicate that as more smart electronics start to emerge, the
110 amount of both PCB and gold that are recoverable from waste electronics could start to increase
111 within the next decade.

112 Due to limited publicly-available tear-down analyses, the modeled PCB and gold content
113 assumed in different types of electronic devices has a relatively large range (Figs. 2c and 2d).
114 Thus, the impact of composition is much higher than that of the growth scenario, which further
115 affects the economic potentials analyzed in the discussion section.

116 In terms of zip-code level distributions in the U.S., as the model is developed based on the
117 population and household possession data (See *Methods* and Supplementary Note 4)²⁶⁻²⁸, the
118 waste electronics densities are heavier in the coastal and metropolitan areas (Fig. 3a). It is notable
119 that in the densest regions (i.e., certain places New York City and Los Angles), roughly over
120 160,000 kg of waste electronics could be generated annually within one square kilometer.

121 For metal extraction, the mining and refining sector is an established industry in the U.S. that can
122 potentially be utilized to treat waste electronics or their embedded PCBs, as both involve hydro or
123 pyrometallurgical processes.⁷ In terms of the treatment process, after the waste PCBs are
124 dismantled, shredded, and physically separated into metal scraps (upstream recycling), the
125 metallurgical processes (downstream recycling) are fairly similar whether refining metal from
126 these waste or virgin mines.^{29,30} In fact, a majority of the metal recovery from waste electronics is
127 still based on pyrometallurgical pathways, which is also the primary virgin mining process with a
128 minor difference between feedstocks (i.e., precious metal vs. base metal refining).³¹⁻³³

129 Virgin gold mines and refining plants mostly exist in the western states (i.e., Nevada and
130 Arizona) of the U.S., as shown in Fig. 3b. The rest of the virgin gold mines are scattered around
131 the U.S., but relatively more concentrated in the mountain areas (i.e., Utah and Colorado). Again,
132 before the waste electronics could be processed in metal refinery plants, they would need to be
133 collected and pre-processed (i.e., sorting shredding, and physical separation) by different
134 recyclers.

135 Collection of waste electronics from consumers is out of the scope of this study, but Fig. 3b
136 shows an estimate of the locations of certified waste electronics recyclers as recognized by the
137 USEPA³⁴ (green dots), and their relationship to the current virgin gold mines and/or refineries
138 (triangles). Locations of the certified waste electronics recyclers generally show good agreement
139 with the distribution of waste electronics generated in Fig. 3a and also indicate that there are
140 unique challenges and opportunities for handling waste electronics among the different regions in
141 the U.S. For example, there could be a potential overlap between virgin mining production and
142 the potential metal recovery from waste electronics mostly in the western part of the U.S.,
143 whereas there could be more potential for building new waste electronic mining facilities in the
144 Central or Eastern regions.

145 *Growth of different types of waste electronics*

146 This study shows that currently, the U.S. is experiencing a dramatic change in waste electronics
147 composition. The total amount of small-to-midsize consumer waste electronics and the potential
148 precious metal (gold) recoverable from that waste has been steady in recent years, however, our
149 model suggests that the value of waste electronics will likely increase in the next decade owing to
150 the changes in metal composition and shipments from emerging electronics. Specifically, Fig. 4a
151 shows the predicted growth of representative waste electronics in 2021-2022 and 2032-2033 (See
152 the whole list in Supplementary Dataset). Most of the waste electronics growing in 2033 also
153 have positive growth rates in 2022. The overall scale of the difference in growth rates in 2022 is
154 much larger (up to over 400%) than that of 2033 (up to 30%) owing to one or two recent spikes
155 between 2018-2021, whose effect is diminished when projected to 2033. Notably, several of the
156 waste from several emerging types of electronics, such as 5G smartphones and gateways, electric
157 scooters, and wireless earbuds have large range of predicted growth rates due to high sales
158 fluctuations in recent years. For these electronics, further market analysis is recommended to
159 provide more accurate growth predictions. Also, note that for the future scenario (2032-2033), the
160 MFA model prediction in this study is a conservative estimate because this study does not
161 account for completely new electronics that might enter the market with potentially high growth
162 rates.

163 Furthermore, emerging electronics such as AR/VR, connected home devices, and internet of
164 things products (defined as electronics requiring constant wireless communications) are leading
165 the growth rate in 2032-2033. Other waste smart electronics (i.e., wireless headphones, drones)
166 did not make it to the top 10 growing list but also have close to, or higher than 10% growth rate
167 from 2032 to 2033. On the other hand, most of the conventional waste electronics (i.e., desktops
168 PCs, printers, DVDs) are generated at a steady rate or declining. These dramatic differences in the
169 growth patterns between different waste electronics show that the U.S. is currently experiencing a

170 shift in the composition of waste electronics, towards smaller, portable, and more complex
171 electronics, which potentially contain more resource value in terms of weight percentage²⁵.
172 In terms of mass generated, Figs. 4 (b and c) show that certain heavy electronics such as LCD
173 TVs greatly contribute to the overall mass of waste electronics, similar to how CRT TVs
174 dominated the mass of waste electronics in the early 21st century.²¹ Other heavy electronics, such
175 as printers and desktop PCs are still among the top ten waste electronics by mass. Lastly, smaller
176 computers such as laptops and tablets will likely exceed conventional desktops or monitors in the
177 waste stream by 2033.

178

179 **Discussion**

180 *Economic Potential*

181 The geospatial modeling shows that there are potential underlying connections and opportunities
182 between virgin mining and recycling in the U.S., which can be used to help create a circular
183 economy around metal recovery from waste electronics. By evaluating the capacity and
184 profitability of the virgin mining refineries, we find that it is worthwhile considering the
185 integrated pathway of recycled waste electronics and virgin metal recovery routes in the U.S.
186 Since gold can represent over 85% of the embedded value within consumer electronics¹³⁻¹⁵, it is
187 used as the primary indicator to study the direct economic potential of treating electronics. One
188 important criterion is to compare the potential gold productivity from the waste electronics with
189 the current gold productivity throughout the U.S., which is approximately 220 tonnes annually³⁵.
190 Fig 5a shows that the gold recoverable from the waste electronics can potentially reach the same
191 magnitude as the national productivity from virgin resources when assuming gold compositions
192 within the electronics are on the higher end of values found in the literature.^{21,24,25,36}

193 Theoretically, the virgin mining refining plants have the capacity of to handle the total quantity of
194 waste electronics for precious metal (gold) recovery. However, after considering the geological
195 distributions and current gold mining production, virgin mining with large handling capacities for
196 waste electronics is concentrated in the West and Mountain areas of the U.S. The Central and
197 East regions may have more need to create new recycling infrastructure targeting waste
198 electronics.

199 To elaborate, on the state level (estimated in Supplementary Note 5), Nevada and Alaska are the
200 leading states for virgin gold production (Fig. 5b), with capabilities of approximately 173 metric
201 tonnes and 21 metric tonnes per year (2018 value published in 2021), respectively.³⁵ Due to its
202 high virgin gold productivity, refining plants at mines in Nevada should theoretically have the
203 capacity necessary to handle all of the waste electronics in the U.S. but would face extra burdens
204 (i.e., time, cost, emissions) when transporting waste electronics that are generated far from the
205 region.

206 If the embedded gold from waste electronics in certain regions reaches the maximum allocated
207 productivity (light blue color), it would need to be transported to the next nearest facility with
208 excess productivity (dark blue color). In this case, the transportation burden increases with the
209 distance, which is qualitatively represented by the darkness of the yellow color in Figs. 5(c-h),

210 In recent years, there have been emerging research and proposals on new technologies to recover
211 metals from waste electronics, such as electrochemical treatment, supercritical aided extraction,
212 photocatalysis, bioleaching, etc.^{7,31,37,38} In the near future, it is possible that more efforts will be
213 made to commercialize these technologies specifically to recover resources from waste
214 electronics treatment. These new facilities may need to compete with virgin mining refineries that
215 integrate electronic waste recycling into their production capacity. Thus, a higher transportation
216 burden to integrated virgin refineries indicates greater potential to build such new facilities
217 (represented by dark yellow color). In Figs. 5(c-h), the darker blue color represents a higher

218 opportunity to integrate waste electronics with existing virgin mining refineries, whereas the
219 darker yellow color represents the higher economic potential for new facilities.

220 The MFA results (Fig. 3) and Fig. 5a show that the uncertainties in the embedded gold content
221 have a much higher influence compared with the growth scenario, therefore are chosen as one of
222 the main parameters for sensitivity to the recycling economic potential. Other key parameters
223 include first, the level of involvement, which denotes how much productivity can the virgin
224 refineries allocate to use waste electronics as feedstock. Second, the influence of exportation is
225 also studied, as it directly relates to the level of domestic recycling of waste electronics. We have
226 found that the reported degree of waste electronics exportation in the U.S. has high variation
227 depending on the time of study and reporting agencies.^{20,39}

228 Fig. 5 shows that within the uncertainties studied, the economic potential is most sensitive to the
229 metal content, as shown when comparing Figs. 5(c-e) with Figs. 5(f-h). In comparison, the level
230 of exportation and involvement of the current virgin mining industry will affect the East Central
231 regions if the embedded values are low, as shown in Figs. 5(f-h). Figs. 5(c-e) show that if the
232 embedded metal content is high, there is high economic potential to develop new infrastructure
233 around waste electronics recycling even if there is high exportation, or if the virgin mines shift a
234 large portion (up to around 70%) of their productivity to be generated from waste electronics.

235 It is important to note that metal compositions are different between waste electronics and virgin
236 mines,^{36,40} which would require more separation stages. Results from this section imply that
237 quantifying the economic trade-off between the exact procedures that need to be added for the
238 virgin mines to handle electronics can make a significant difference in examining the nationwide
239 profitability. Thus for this purpose, future research is recommended on the techno-economic
240 comparison between adapting virgin metal refineries, particularly gold, to include separated
241 electronics as part of the feed stream, versus building completely new plants.

242

243 Future Outlooks

244 The above discussions highlight potential opportunities to enhance the circular economic
245 potential of metal recovery from waste electronics recycling and help address several major
246 challenges identified in this study. Future research should focus on process development for both
247 upstream (recyclers and collectors) and downstream refineries for this integrated waste and virgin
248 mining pathway to enhance the economic potential. Additionally, policy efforts should be focused
249 on creating a national-level database that includes the composition-level tear-down data, location
250 of small-scale refineries, and collection plants for various electronics to help narrow the range for
251 future MFAs and offer a more complete geospatial analysis of metal recovery from waste
252 electronics.

253 First and most importantly, a better understanding of metal compositions within different types of
254 electronics is needed, especially among emerging smart electronics. The availability range plotted
255 in Fig. 2 and Fig. 5 is relatively large due to the fact that limited tear-down studies are available
256 for waste electronics. Between the two main uncertainties included in this analysis, the
257 composition has a significantly higher impact on the available gold within the waste electronics
258 than the growth scenarios.

259 The large range of resource availability indicates that different management strategies might be
260 needed when aiming to create a circular economy around waste electronics. A database that
261 includes the compositions of metals for different types of electronics would help anticipate future
262 recycling infrastructure needs. Such a database could be achieved via high-quality tear-down
263 analysis, as well as help from the electronics manufacturers without exposing company
264 intellectual properties.

265 Second, there has been limited transparency at the national level on both the upstream recyclers
266 (green dots in Fig. 3b) and metal refineries (triangles in Fig. 3b). There are many waste
267 electronics recyclers and refineries (i.e., small-scale and/or regional certified facilities) with

268 minimal publicly available information and transparency that may require a national-level survey
269 or reporting database.

270 Furthermore, although gold is chosen as the primary driven factor for economics due to its high
271 embedded cost values, other resources in waste electronics can also add value to the circular
272 economic potential of waste electronics recycling. For example, plastics in waste electronics,
273 which can occupy up to 30 wt%,⁴¹ could be recycled for re-manufacture or energy conversion
274 purposes⁴². However, it is important to note that certain detoxification procedures might be
275 required to eliminate the effects of brominated flame retardants during the high-temperature
276 treatments.^{31,43} Other metals, such as copper and aluminum, are not as valuable as gold but are
277 also important to other manufacturing industries. More importantly, rare earth elements in
278 magnetics and PCBs, and cobalt and lithium in batteries, can help with increasing the supply
279 chain resiliency of critical materials^{44,45} and should be considered in future studies.

280 Lastly, note that since this study does not include all of the consumer or non-consuming waste
281 electronics, the economic potential and profitability is a relatively conservative estimate. If other
282 sources of waste electronics (i.e., the growing electronics in vehicles and industrial plants) are
283 included, the yellow portion of the profitability maps can potentially be expanded.

284

285

286

287

288 **Methods**

289 Material Flow Analysis

290 Compared with the previous MFA models for waste electronics,¹⁹⁻²¹ we expanded the types of
291 waste electronics from the 20 common electronics (i.e., waste cell phones, TVs) to over 90
292 different types of electronic devices. The waste electronics covered in this study included
293 emerging waste electronics such as wearable fitness and health products, portable and wireless
294 devices, smart home improvement devices, AR/VR sets, and consumer drones. The sales data
295 from the Consumer Technology Association (CTA) was used.⁴⁶ The types of electronics modeled
296 in this study were included in the supplementary dataset.

297 The sales data of the target consumer electronics after 2021 was predicted via a logistic-based
298 model, which is known to be capable of capturing the market behavior of consumer products.⁴⁷⁻⁴⁹
299 The logistic model could be described by equations 1 and 2.

300 $sales = \frac{\partial}{1+e^{\beta(t-\gamma)}} \quad (1)$

301 $sales = \frac{\partial}{1+e^{-\beta(t-\gamma)}} \quad (2)$

302 where t denoted the number of years for market sales; ∂ (saturation), β (steepness), and γ
303 (midpoint), were the logistic parameters. Note that Equation 1 was used for the declining
304 electronics, and Equation 2 was used for the growing/increasing electronics.

305 To account for the robust change in consumer behaviors, the last-reported 3 years of sales data for
306 different electronics were first used to categorize the electronics into “decreasing” or “increasing”
307 patterns. For the electronics that had decreasing sales patterns, we compared the logistic fitting
308 since the maximum reported value and the 5 most-recent values, and selected the one with the
309 larger r square value to predict the future sales data. The same method was used to predict the
310 sales data beyond the reported years if the reporting stopped before 2021. For the increasing

311 electronics, we set the defined “fast growth” and “slow growth” scenarios using logistic fitting
312 based on different assumptions of maximum sales penetration per household values. More details
313 on categorizing the growth scenarios and sales data prediction were included in Supplementary
314 Note 1.

315 A Weibull probability function was used to predict the temporal evolution of waste electronics in
316 the MFA model, which has been used in previous studies to predict the flow of waste electronics
317 based on their life span.^{21,50} In this model, we assumed that the probability of a certain type of
318 electronic device (j) reaching its end of life within its maximum usage year (n) followed a
319 Weibull distribution, which could be described by Equations (3) and (4)¹⁹⁻²¹.

$$320 \quad f(\eta, \delta, t) = \frac{\delta}{\eta} \left(\frac{t}{\eta}\right)^{(\delta-1)} e^{-\left(\frac{t}{\eta}\right)^\delta} \quad (3)$$

321 In the probability density function (3), t represented time, and δ and η were the parameters used
322 to describe the Weibull probably function. The probability of reaching end-of-life (P) could be
323 then used, along with the historical sales data (S) to calculate the flow (N) of waste electronics j
324 generated after a lifespan of i years within its maximum life span (n).²¹

$$325 \quad N_{j,t} = \sum_{i=1}^n P_{j,i} \times S_{j,t-i} \quad (4)$$

326 The above-mentioned MFA model was applied to the U.S. shipment data of various electronics
327 evaluated in this study. The model is conducted using the open-source MATLAB code as
328 described in Althaf et al. 2021,²¹ with adjustments to the numbers and types of electronics,
329 Weibull parameters, average mass, compositions, etc.

330 For the common electronics that were analyzed in previous studies (i.e., waste cell phones, TVs),
331 we combined the previously-published Weibull parameters and life span from different
332 sources.^{19,21} It was noted that the sum of probabilities function derived from the previous Weibull
333 parameters is slightly less than 1 if added up to the same reported maximum life span in previous
334 literature, in which the deviation probability was assumed to be the devices that can be further

335 reused when predicting the amount of waste generated. But the reused or refurbished electronics
336 were not included for further analysis (also included later in *Key assumptions and uncertainties*).
337 For the relatively sparsely-studied electronics (i.e., smart electronics, wireless electronics, AR/VR
338 sets), their MFA parameters were determined from those of the 20 common electronics based on
339 UNU classification codes, Harmonized System (HS) trade codes, and functionalities.^{51,52} Details
340 of the decision tree and sources for determining the MFA parameters are included in
341 Supplementary Notes 2 and 3. The material compositions of different waste electronics were
342 determined from previous literature. The upper and lower bounds of the reported values were
343 used as the “high content” and “low content” scenarios, respectively, with details shown in
344 Supplementary Note 3.

345 Geospatial Modeling

346 To model the spatial distribution of the waste electronics generated across the U.S., the total MFA
347 results were distributed to each zip code area based on the population density and amount of
348 electronics in residential households and commercial office buildings within different geological
349 regions of the U.S.(i.e., New England, Pacific, etc.)^{27,28} Regional results were normalized to waste
350 electronics generated per capita for different zip code areas with the region.

351 The average percentage of ownership across different regions of the U.S. were calculated based
352 on the possession and ownership data for sixteen types of electronics (i.e., smartphones, TVs, cell
353 phones, desktop, and laptop computers) in residential households and five types of electronics in
354 commercial office buildings provided by the U.S. Energy Information Administration
355 (USEIA)^{27,28}. The total amount of waste electronics generated from the MFA was corrected and
356 distributed to different regions according to their average percentage of ownership and divided by
357 their total population to obtain the average generation/capita for different regions. Note that due
358 to this assumption, the broader distribution of ownership per household of various electronics
359 might be different than the representative types of electronics used.

360 Waste electronics generated per capita for each geological region were multiplied by the
361 population density data for each zip code in the U.S. to estimate the waste electronics generated
362 for each zip code. The waste electronics generated per zip code were combined with
363 corresponding geospatial data (shape, boundary, longitude, latitudes, etc.) to plot the distribution
364 of the U.S. Nation-wide geological shape data used in this study was obtained from the U.S.
365 Census Bureau.⁵³

366 To assess the potential connections between waste electronics recycling and virgin refineries, the
367 geospatial coordinates for the mines, mining plants, and their state-level productivities in the U.S.
368 were obtained from various U.S. Geological Survey (USGS) sources.^{35,54-56} The nationwide
369 certified recycler data was provided by Sustainable Electronics Recycling International (SERI)⁵⁷
370 based on their certified recycler lists across the U.S.

371 Note that besides virgin mining, there are also several existing major refineries that list waste
372 electronics as part of the feedstocks (See Supplementary Table S4) to produce high-quality metals
373 for the technology industry.^{58,59} As there is limited information on the gold productivity and
374 feedstock composition for these refineries, we estimated their influence on the economic potential
375 in Supplementary Note 6, by comparing with virgin plants' productivities, previous techno-
376 economic analysis, and tear-down studies.^{13,14,21,60}

377 Further, we apply a distance matrix to first determine the potential capability for treating waste
378 electronics at the nearest existing facility, either through integrating with virgin mining refineries.
379 We assume that these facilities can allocate/expand a certain portion of their current gold
380 productivity to waste electronics. We further characterize the U.S. into five greater regions shown
381 in the legend of Fig. 5 and analyzes their handling capabilities in Figs. 5(c-h). More details on the
382 geospatial modeling, and productivity and profitability analyses were included in Supplementary
383 Notes 4 to 6.

384 Key Assumptions and Uncertainties

385 This section summarizes the key source of uncertainties and assumptions used in this study. All
386 of the uncertainties and assumptions were also discussed in the previous text or the
387 Supplementary Notes 1-6 when they were applied to the corresponding analysis. To summarize,
388 the key assumptions and sources of uncertainties in this study were summarized as follows. First,
389 the scope of this study includes most of the mid to small-size consumer electronics. Due to this
390 assumption, we expect that the overall amount of waste electronics will be higher than what is
391 predicted in this study. Second, refurbished or reused electronics are not modeled in this study.
392 As described in the *Material flow analysis* Section, the re-use was recognized by assuming not all
393 of the electronics sales reach end-of-life in the probability distribution based on the reported
394 Weibull parameters and maximum life spans, but their further re-introduction to the waste stream
395 was not included in this analysis.

396 For key uncertainties considered in this study, first, due to limited data availability, uncertainties
397 in this study included the composition of resources within the electronics wastes, degree of
398 domestic recycling (as studied by the percent export), approximation of MFA results, and spatial
399 analysis by using representative values. Second, we recognize that there is a high potential that
400 new types of small-to-midsize consumer electronics will be introduced in the near future, similar
401 to how AR/VR and 5G devices have emerged in recent years. This was accounted for by
402 qualitatively showing the possible markup of results in Fig. 2. Lastly, as discussed in the *Future*
403 *Outlook* Section, although the effects of the key sources of uncertainties were studied (namely the
404 growth scenario, content of resources, and level of export) in the *Results* and *Discussion* sections,
405 future research would greatly benefit from narrowing these ranges via more comprehensive tear
406 down data, and improved policy incentives as stated in the *Discussion* Section.

407

408 **Data Availability**

409 The data that support the findings of this study are available within the paper and its
410 Supplementary Information. Supplementary dataset is available at [https://github.com/ppeng-](https://github.com/ppeng-cloud/Consumer-Electronics-Recycling-Potential-in-United-States)
411 [cloud/Consumer-Electronics-Recycling-Potential-in-United-States](https://github.com/ppeng-cloud/Consumer-Electronics-Recycling-Potential-in-United-States).

412 **Code Availability**

413 All steps used in this analysis are illustrated in the Methods section and Supplementary Notes 1-6.
414 Supplementary scripts are available at [https://github.com/ppeng-cloud/Consumer-Electronics-](https://github.com/ppeng-cloud/Consumer-Electronics-Recycling-Potential-in-United-States)
415 [Recycling-Potential-in-United-States](https://github.com/ppeng-cloud/Consumer-Electronics-Recycling-Potential-in-United-States) and from the corresponding author on reasonable request.

416

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427

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429 **Figure Legends/Captions**

430 **Fig. 1. Scope and system boundary of this study.** Pie chart denotes the general distribution of
431 values between gold (yellow) and other metals extracted from metal recycling processes of waste
432 electronics, based on reported values from refs^{14,23}. The value distribution will vary between
433 different electronics products such as type, brand, manufacturer, etc.

434

435 **Fig. 2. Temporal changes for waste electronics generation between 2015-2033 and its**
436 **geospatial distribution results. a.** Number of units generated. **b.** Mass of waste electronics
437 generated. **c.** Mass of gold (most valuable resource) potentially recoverable from waste
438 electronics. **d.** Printed circuit board (PCB) potentially available within waste electronics.
439 Uncertainties are caused by the assumed fast and slow growth scenarios discussed in
440 Supplementary Note 1 (**a** and **b**). Additional uncertainties are included for **c** and **d** when assuming
441 different compositions within the electronics (See Supplementary Notes 1 and 3). Gradient color
442 qualitatively addresses the possibility of increasing in the future after introducing new electronics.

443

444 **Fig. 3. Spatial distribution of waste electronics resources, certified recyclers, and major**
445 **mining plants in the U.S. a.** Estimated relative density map of waste electronics generated per
446 zip code in the U.S. **b.** Modelled spatial distribution of waste electronics and their corresponding
447 recyclers as well as the locations for virgin mining plants within continental U.S..^{53-55,57} See
448 detailed geospatial distribution of mines in Supplementary Note 4. The maps were created in
449 Python v3.8.5 using GeoPandas v0.8.1 (<https://geopandas.org>) using the shape file from the U.S.
450 Census Bureau⁵³

451

452 **Fig. 4. The growth of representative types of waste electronics. a.** The average growth rate of
453 representative electronics between 2032-2033 (dots) and their corresponding growth rate in 2021-
454 2022 (lines). The growth of 5G Smartphones from 2021 to 2022 exceeds 400% and is not shown
455 in the figure. Full growth data available in the Supplementary dataset. **b.** Modelled waste
456 electronics with the largest weight percentage in 2021. **c.** Modelled waste electronics with the
457 highest weight percentage in 2033.

458

459 **Fig. 5. Modelled distribution of valuable (gold) from waste electronics in 2033 and areas**
460 **where the generated waste can be handled by virgin gold plants for gold extraction. a.**
461 Relative amount of gold recoverable in 2033 compared with recent national virgin production.
462 Error bars denote uncertainties in growth scenarios **b.** Relative state-level productivity of gold
463 from virgin sources in the U.S. and map showing where waste electronics are handled based on
464 closest-distance approximation.. **c-h.** Evaluation of the potential to integrate waste electronics to
465 virgin mining if the plant allocate/expand 30% of the plants current productivity (d, e, g, h), or
466 70% of the plants current productivity (c and f) to producing gold from waste electronics. High
467 circularity (c, d, f, g) means 80% of the waste electronics are recycled domestically, and high
468 export (e and h) refers to only 40% of the waste electronics are recycled domestically. Light blue
469 color represents the maximum area these plants are able to cover. Dark blue means that the plants
470 can utilize all of the waste electronics with excess capacity left. Yellow denotes the region where
471 gold productivity from waste electronics is beyond the assumed capacity of the plants. The maps
472 were created in Python v3.8.5 using GeoPandas v0.8.1 (<https://geopandas.org>) using the shape
473 file from the U.S. Census Bureau⁵³

474

475 **References**

- 476 1 Robinson, B. H. E-waste: an assessment of global production and environmental impacts.
477 *Science of the total environment* **408**, 183-191 (2009).
- 478 2 Fiore, S., Ibanescu, D., Teodosiu, C. & Ronco, A. Improving waste electric and
479 electronic equipment management at full-scale by using material flow analysis and life
480 cycle assessment. *Science of The Total Environment* **659**, 928-939 (2019).
- 481 3 Forti, V., Balde, C. P., Kuehr, R. & Bel, G. The Global E-waste Monitor 2020:
482 Quantities, flows and the circular economy potential. (2020).
- 483 4 USEPA. Advancing Sustainable Materials Management: 2014 Fact Sheet. (2016).
- 484 5 Awasthi, A. K., Li, J., Koh, L. & Ogunseitan, O. A. Circular economy and electronic
485 waste. *Nature Electronics* **2**, 86-89 (2019).
- 486 6 Zabala, A. Illegal electronic waste recycling trends. *Nature Sustainability* **2**, 353-354
487 (2019).
- 488 7 Hsu, E., Barmak, K., West, A. C. & Park, A.-H. A. Advancements in the treatment and
489 processing of electronic waste with sustainability: a review of metal extraction and
490 recovery technologies. *Green chemistry* **21**, 919-936 (2019).
- 491 8 Nithya, R., Sivasankari, C. & Thirunavukkarasu, A. Electronic waste generation,
492 regulation and metal recovery: a review. *Environmental Chemistry Letters* **19**, 1347-1368
493 (2021).
- 494 9 Sun, R. *et al.* Bioaccumulation of short chain chlorinated paraffins in a typical freshwater
495 food web contaminated by e-waste in south china: Bioaccumulation factors, tissue
496 distribution, and trophic transfer. *Environmental Pollution* **222**, 165-174 (2017).
- 497 10 Kyere, V. N. *et al.* Contamination and health risk assessment of exposure to heavy metals
498 in soils from informal e-waste recycling site in Ghana. *Emerging Science Journal* **2**, 428-
499 436 (2018).
- 500 11 Purushothaman, M., Inamdar, M. G. & Muthunarayanan, V. Socio-economic impact of
501 the e-waste pollution in India. *Materials Today: Proceedings* **37**, 280-283 (2021).
- 502 12 Palmieri, R., Bonifazi, G. & Serranti, S. Recycling-oriented characterization of plastic
503 frames and printed circuit boards from mobile phones by electronic and chemical
504 imaging. *Waste Management* **34**, 2120-2130 (2014).
- 505 13 Ghodrati, M., Rhamdhani, M. A., Brooks, G., Masood, S. & Corder, G. Techno economic
506 analysis of electronic waste processing through black copper smelting route. *Journal of*
507 *Cleaner Production* **126**, 178-190 (2016).
- 508 14 Diaz, L. A. & Lister, T. E. Economic evaluation of an electrochemical process for the
509 recovery of metals from electronic waste. *Waste Management* **74**, 384-392 (2018).
- 510 15 Patil, T. & Patil, S. T. Techno-economic Feasibility of Recycling E-waste to Recover
511 Precious Metals. *International Journal of Advanced Scientific and Technical Research* **7**
512 (2015).
- 513 16 Islam, M. T. & Huda, N. Material flow analysis (MFA) as a strategic tool in E-waste
514 management: Applications, trends and future directions. *Journal of Environmental*
515 *Management* **244**, 344-361 (2019).
- 516 17 De Meester, S., Nachtergaele, P., Debaveye, S., Vos, P. & Dewulf, J. Using material flow
517 analysis and life cycle assessment in decision support: A case study on WEEE
518 valorization in Belgium. *Resources, Conservation and Recycling* **142**, 1-9 (2019).
- 519 18 Islam, M. T. & Huda, N. E-waste in Australia: generation estimation and untapped
520 material recovery and revenue potential. *Journal of Cleaner Production* **237**, 117787
521 (2019).
- 522 19 USEPA. *Electronic Products Generation and Recycling in the United States, 2013 and*
523 *2014, Office of Resource Conservation and Recovery.* (2016).

524 20 Duan, H., Miller, T. R., Gregory, J., Kirchain, R. & Linnell, J. *Quantitative*
525 *characterization of domestic and transboundary flows of used electronics: Analysis of*
526 *Generation, Collection, and Export in the United States*. 121 (2013).

527 21 Althaf, S., Babbitt, C. W. & Chen, R. The evolution of consumer electronic waste in the
528 United States. *Journal of Industrial Ecology* **25**, 693–706 (2021).

529 22 Duman, G. M., Kongar, E. & Gupta, S. M. Estimation of electronic waste using
530 optimized multivariate grey models. *Waste Management* **95**, 241-249 (2019).

531 23 Golev, A., Corder, G. D. & Rhamdhani, M. A. Estimating flows and metal recovery
532 values of waste printed circuit boards in Australian e-waste. *Minerals Engineering* **137**,
533 171-176 (2019).

534 24 Golev, A., Schmeda-Lopez, D. R., Smart, S. K., Corder, G. D. & McFarland, E. W.
535 Where next on e-waste in Australia? *Waste management* **58**, 348-358 (2016).

536 25 Babbitt, C. W., Madaka, H., Althaf, S., Kasulaitis, B. & Ryen, E. G. Disassembly-based
537 bill of materials data for consumer electronic products. *Scientific Data* **7**, 1-8 (2020).

538 26 *Historical Population Change Data (1910-2020)*, US Census Bureau, Accessed
539 07/01/2021, [https://www.census.gov/data/tables/time-series/dec/popchange-data-](https://www.census.gov/data/tables/time-series/dec/popchange-data-text.html)
540 [text.html](https://www.census.gov/data/tables/time-series/dec/popchange-data-text.html).

541 27 *2020 RECS Survey Data*, U.S. Energy Information Administration, Accessed 07/04/2021,
542 <https://www.eia.gov/consumption/residential/data/2020/>.

543 28 *2018 CBECS Survey Data*, U.S. Energy Information Administration, Accessed
544 07/01/2022,
545 <https://www.eia.gov/consumption/commercial/data/2018/index.php?view=microdata>.

546 29 Ghimire, H. & Ariya, P. A. E-wastes: bridging the knowledge gaps in global production
547 budgets, composition, recycling and sustainability implications. *Sustainable Chemistry* **1**,
548 154-182 (2020).

549 30 Tabelin, C. B. *et al.* Copper and critical metals production from porphyry ores and E-
550 wastes: A review of resource availability, processing/recycling challenges, socio-
551 environmental aspects, and sustainability issues. *Resources, Conservation and Recycling*
552 **170**, 105610 (2021).

553 31 Peng, P. & Park, A.-H. A. Supercritical CO₂-induced alteration of a polymer–metal
554 matrix and selective extraction of valuable metals from waste printed circuit boards.
555 *Green Chemistry* **22**, 7080-7092 (2020).

556 32 Kaya, M. Recovery of metals and nonmetals from electronic waste by physical and
557 chemical recycling processes. *Waste management* **57**, 64-90 (2016).

558 33 Wang, H. *et al.* Recovery of waste printed circuit boards through pyrometallurgical
559 processing: A review. *Resources, Conservation and Recycling* **126**, 209-218 (2017).

560 34 *Certified Electronics Recyclers*, United States Environmental Protection Agency,
561 Accessed 02/24/2020, [https://www.epa.gov/smm-electronics/certified-electronics-](https://www.epa.gov/smm-electronics/certified-electronics-recyclers)
562 [recyclers](https://www.epa.gov/smm-electronics/certified-electronics-recyclers).

563 35 USGS. *Minerals Yearbook - Gold*. (2021).

564 36 Priya, A. & Hait, S. Comprehensive characterization of printed circuit boards of various
565 end-of-life electrical and electronic equipment for beneficiation investigation. *Waste*
566 *Management* **75**, 103-123 (2018).

567 37 Chen, Y. *et al.* Selective recovery of precious metals through photocatalysis. *Nature*
568 *Sustainability* **4**, 618-626 (2021).

569 38 Uekert, T., Pichler, C. M., Schubert, T. & Reisner, E. Solar-driven reforming of solid
570 waste for a sustainable future. *Nature Sustainability* **4**, 383-391 (2021).

571 39 Işıldar, A., Rene, E. R., van Hullebusch, E. D. & Lens, P. N. L. Electronic waste as a
572 secondary source of critical metals: Management and recovery technologies. *Resources,*
573 *Conservation and Recycling* **135**, 296-312 (2018).

574 40 Jones, R. S. & Fleischer, M. *Gold in minerals and the composition of native gold*. 2330-
575 5703. ([US Dept. of the Interior, Geological Survey], 1969).

576 41 Riise, B. in *Energy Technology 2020: Recycling, Carbon Dioxide Management, and*
577 *Other Technologies* 295-305).

578 42 Heller, M. C., Mazor, M. H. & Keoleian, G. A. Plastics in the US: toward a material flow
579 characterization of production, markets and end of life. *Environmental Research Letters*
580 **15**, 094034 (2020).

581 43 Chien, Y.-C., Paul Wang, H., Lin, K.-S., Huang, Y. J. & Yang, Y. W. Fate of bromine in
582 pyrolysis of printed circuit board wastes. *Chemosphere* **40**, 383-387 (2000).

583 44 Dushyantha, N. *et al.* The story of rare earth elements (REEs): Occurrences, global
584 distribution, genesis, geology, mineralogy and global production. *Ore Geology Reviews*
585 **122**, 103521 (2020).

586 45 Godoy León, M. F., Matos, C. T., Georgitzikis, K., Mathieux, F. & Dewulf, J. Material
587 system analysis: Functional and nonfunctional cobalt in the EU, 2012–2016. *Journal of*
588 *Industrial Ecology* (2022).

589 46 *January 2021 FastFacts Historical Sales Data*, Consumer Technology Association
590 (CTA), Accessed 09/19/2021, <https://shop.cta.tech/collections/research>.

591 47 Müller, E., Hilty, L. M., Widmer, R., Schluep, M. & Faulstich, M. Modeling metal stocks
592 and flows: a review of dynamic material flow analysis methods. *Environmental science*
593 *& technology* **48**, 2102-2113 (2014).

594 48 Althaf, S., Babbitt, C. W. & Chen, R. Forecasting electronic waste flows for effective
595 circular economy planning. *Resources, Conservation and Recycling* **151**, 104362 (2019).

596 49 Liu, X., Tanaka, M. & Matsui, Y. Generation amount prediction and material flow
597 analysis of electronic waste: a case study in Beijing, China. *Waste management &*
598 *research* **24**, 434-445 (2006).

599 50 Gu, Y., Wu, Y., Xu, M., Mu, X. & Zuo, T. Waste electrical and electronic equipment
600 (WEEE) recycling for a sustainable resource supply in the electronics industry in China.
601 *Journal of Cleaner Production* **127**, 331-338 (2016).

602 51 Forti, V., Baldé, K. & Kuehr, R. *E-waste statistics: guidelines on classifications,*
603 *reporting and indicators*. (United Nations University, 2018).

604 52 *Harmonized System (HS) Codes*, Accessed 07/06/2021,
605 [https://www.trade.gov/harmonized-system-hs-](https://www.trade.gov/harmonized-system-hs-codes#:~:text=The%20United%20States%20uses%20a,Census%20Bureau%27s%20Fore)
606 [codes#:~:text=The%20United%20States%20uses%20a,Census%20Bureau%27s%20Fore](https://www.trade.gov/harmonized-system-hs-codes#:~:text=The%20United%20States%20uses%20a,Census%20Bureau%27s%20Fore)
607 [ign%20Trade%20Division](https://www.trade.gov/harmonized-system-hs-codes#:~:text=The%20United%20States%20uses%20a,Census%20Bureau%27s%20Fore).

608 53 United States Census Bureau, Accessed 01/31/2021, <https://www.census.gov/data.html>

609 54 USGS. U.S. Geological Survey, Active Mines and Mineral Processing Plants in the
610 United States in 2003, U.S. Geological Survey (USGS), Reston, Virginia., (2005).

611 55 Mining Data Online, Accessed 02/24/2021,
612 <https://miningdataonline.com/property/list.aspx?vw=3>.

613 56 Sheaffer, K. N. *Gold Data Sheet - Mineral Commodity Summaries 2020*. 70-71 (2020).

614 57 *Find an R2 Certified Facility*, SERI (Sustainable Electronics Recycling International),
615 Accessed 04/01/2021, <https://sustainableelectronics.org/find-an-r2-certified-facility/>.

616 58 *Smelter and Refiner List*, Apple Inc., Accessed 01/03/2021,
617 <https://www.apple.com/supplier-responsibility/pdf/Apple-Smelter-and-Refiner-List.pdf>.

618 59 *List of the Smelters or Refiners identified in Konica Minolta's supply chain which were*
619 *known by RMI (As of March 31, 2020)*, Accessed 01/03/2021,
620 <https://www.konicaminolta.com/about/csr/csr/suppliers/pdf/smelters.pdf>.

621 60 Kasper, A. C. & Veit, H. M. Gold recovery from printed circuit boards of mobile phones
622 scraps using a leaching solution alternative to cyanide. *Brazilian Journal of Chemical*
623 *Engineering* **35**, 931-942 (2018).

