Mortality profiles and life-course studies of psychiatric patients populations are far from having lost their epidemiological appeal. We can list the following open questions:

• What is the specific disease profile of each psychiatric condition? Is it suggestive of peculiar patho-physiological processes?
• What are the main co-morbidities or absence of co-morbidities? Are they causally related to treatment?
• Do psychiatric patients suffer inequalities in health? What is the implied web of causation?

The research facing those questions must address major methodological problems: selection bias and reverse causation; time-dependent confounders that are also intermediate variables; complex relationships within a life course.

Psychiatric patients have a higher mortality rate than the general population (Harris & Barraclough, 1998). The excess is greater in hospital-based cohort studies. Meloni et al. (2006) found a threefold mortality risk in a cohort of psychiatric in-patients in Florence. Grigoletti et al. (2009) documented that psychiatric patients in contact with South Verona Community-based Mental Health Service have an almost twofold higher mortality rate.

The risk depends on psychiatric diagnosis, gender and age. It is higher for young females, drug addicts or alcoholists (Amaddeo et al., 1995; Prior et al., 1996). Among the top ranking causes of death we found accidents and suicide, infectious diseases, blood diseases, osteo-articular diseases, genito-urinary diseases, nervous system diseases, respiratory, digestive and cardiovascular diseases. The excess for malignant neoplasm is less prominent. (Roshanaei-Moghaddam & Katon, 2009; Saha et al., 2007).

How can we explain such striking differentials? In fact, the definition of a reference population is far from being straightforward. Psychiatric patients population is composed of people who are mostly excluded from the economically active subset of the general population. Therefore it could be argued that assuming the general population as reference would induce an un-healthy population bias which is symmetrical to the well documented “healthy worker effect” in occupational cohort studies. The “healthy worker effect” depends on the tendency of those working to be selected for their mental and physical ability to work - i.e. being in good health status. For example, economically inactive population shows a relative age-standardized mortality of 3.06 (men) and 2.40 (women) in the 1980’s taking fixed at 1.0 the economically active (Martikainen & Valkonen, 1999). In the studies on the mortality differential by social class, the bias due to the exclusion of economically inactive subset from
the reference population had been discussed. The consequence was an underestimation of the social gap (Blakely & Fawcett, 2005). Here we may have the opposite phenomenon. When studying the mortality differential of psychiatric patients it had been the case for bias due to the inclusion of economically active subset in the reference population. The consequence would be an overestimation of the mortality gap. Notice that the abovementioned all cause mortality relative risks for psychiatric patients range between 2 and 3 with respect to the general population, while those attributable to the economically inactive were between 1.5 and 2 (table 1, page 900, in Martikainen and Valkonen, 1999).

This potential distortion should disappear for longer follow-up times (Goldblatt et al., 1990): in the South Verona study we observed a standardized mortality ratio (SMR) of 3.4 in the follow-up period of ≤1 year from first contact, 2.6 during the second year, 1.8 up to five years, 1.5 up to ten years and 1.4 after ten years (Grigoletti et al., 2009). Two comments: first, even at longer follow-up times there still exists a mortality gap; second, how exclusion from economically active population exerts its ill potential has to be clarified.

Mortality for cancer has a distinct pattern. The excess risk for all malignant neoplasm is not as large as it might be expected (Grinshpoon et al., 2005; Barak et al., 2005). But, noteworthy, oral cavity, lymphohematopoietic and central nervous system tumours are in large excess and tumours of digestive system (stomach), breast, prostate were reported less frequently in comparison with the general reference population (Batul et al., submitted for publication).

Those patterns are specific of each psychiatric diagnosis: alcoholism and psychiatric admission for drug addiction are the conditions stronger associated with unfavourable mortality risk. The more plausible explanation is that those patients have risk factors for other diseases, like infectious, digestive system diseases or oral cancer. This explanation is parsimonious, we do not need additional hypothesis. The mortality differential of psychiatric patients cohorts would be the consequence of higher prevalence of exposure to known risk factors in comparison to the general population. There is evidence that prevalence of smoking is lower in the general (healthier) population than among psychiatric patients (Poirier et al., 2002). If we would be able to define a reference population with equal prevalence of smokers and other life habits then the observed mortality disadvantage of psychiatric case population would disappear.

We can generalize this by means of causal graphs (Pearl, 2009). A graph is composed of nodes (variables) and arrows (direct causal relationships). An indirect causal relationship between two variables will exist whenever there is a direct path connecting the two through other connected nodes (i.e. with head-to-tail arrows). Two variables are associated if there is an open path between them. A path is open if there is no collider on it, i.e. two arrowheads meets on a given node. If we condition on a variable (by stratification or other statistical procedures like standardization and multiple regression) which is not a collider we will block the path; if the variable is a collider, conditioning on it will open the path.

Figure 1 illustrates the problem discussed insofar. The selection indicator S is a collider between P (P=1 denotes being a psychiatric patient) and G (G=1 being member of the active population). Clearly Y (survival) depends on G (population characteristics). Then, conditioning on S, we will observe that P and Y (survival) are associated because there is an open path from P to Y (see Rothman et al., 2008 for a discussion on healthy worker effect, selection bias and confounding).

This explanation of the mortality differential is however partial and unsatisfactory. Partial because excess risk for certain diseases is clearly a reverse causation effect: the underlying pathological condition caused the occurrence of psychiatric symptoms (e.g. organic psychosis). Unsatisfactory because the co-morbidity patterns is not entirely consistent with the toxicology of the involved risk factors.

Leaving aside the still unproven idea that schizophrenic patients have a lower cancer risk (Cohen et al., 2002), the relationship between antipsychotic treatment and co-morbidities received recently great attention (Chwastiak, 2009).

Antipsychotic agents of the second generation can induce a metabolic syndrome and therefore constitute a
risk factor for cardiovascular disorders (Newcomer, 2007). Since they have been widely used for in the last twenty years, psychiatric patients cohorts provided empirical evidence on the risk of mortality for cardiovascular diseases associated to second generation antipsychotic drugs (De Hert et al., 2009). Selection bias may still be an explanation of the observed empirical association, mainly because patients with mental illnesses do have a pre-treatment increased prevalence of risk factors for cardiovascular disease (overweight and obesity, hyperglycemia, dyslipidemia, hypertension, and smoking).

However, adverse metabolic side effects associated with psychotropic medications, such as antipsychotic drugs, were widely documented and several guidelines were diffused (American Diabetes Association et al., 2004). A recent Finnish cohort study (Tiihonen et al., 2009) found a strong heterogeneity of risk among different antipsychotic agents. It is difficult to interpret the comparison with the general population. The study reported a persistent mortality gap. The authors claimed that if a major risk for cardiovascular death be associated to the new generation of antipsychotic drugs it would have been observed an increasing mortality gap. This is a weak argument, indeed, since so many factor and biases may have affected such observation. More interesting, they documented a differential risk among antipsychotic drugs. The statistical analysis used marginal structural models (Robins et al., 2000) addressing potential confounding in the data. Treatment at a given time point depends on previous treatment history, present and past subject conditions. On the other side, we want to assess the causal role of the whole treatment history (cumulative exposure to a given drug) on cardiovascular mortality. The factors which constitute indication for treatment is itself a determinant of the disorder leading to death. In the original paper (Wang et al., 2005) instrumental variable technique was used to make causal inference on the risk associated to antipsychotic drugs.

Second generation antipsychotic drugs raised also a concern when used in non psychiatric patients, elderly patients and patients with dementia (Wang et al., 2005). Several reports documented a higher risk of all cause mortality. Of course the major menace to the validity of this observation is that the underlying condition which constitutes an indication for treatment is itself a determinant of the disorder leading to death. In the original paper (Wang et al., 2005) instrumental variable technique was used to block the indirect path Tk-I-U-Y. A solution proposed by Robbins (2000) is to not condition on I and balance Tk using inverse probability weights proportional to Prob(Tk|Tk-1, I), the probability to receive treatment Tk given the past treatment history and the intermediate confounding variable.

An instrument is a variable which is not associated to the event under study nor it is associated with the confounders but it is a strong predictor of the exposure. A typical instrument is group assignment in a randomized trial. In that study (Wang et al., 2005), the instrument was the physician’s prescribing preference and the analysis used the preferred drug as exposure instead of the drug actually prescribed to the patient.

Let concentrate on Figure 3. The instrument variable Z is strongly associated with treatment T and it has no effect on the outcome (no direct path pointing to Y). The unmeasured variable U will confound the association T-Y (through the open path T-U-Y) but does not confound the association Z-Y (the path Z-T-U-Y is blocked by the collider T).

Figure 2 - A causal diagram with a time-dependent factor which is intermediate for previous treatment and confounder.
These findings underline the need for research on health promotion and preventive programs targeted to psychiatric patients.

Recently, strong emphasis in the literature was dedicated to improve capability to manage medical health problems of psychiatric patients by Mental Health Services and Professionals (McIntyre, 2009; Newcomer, 2009).

**REFERENCES**


